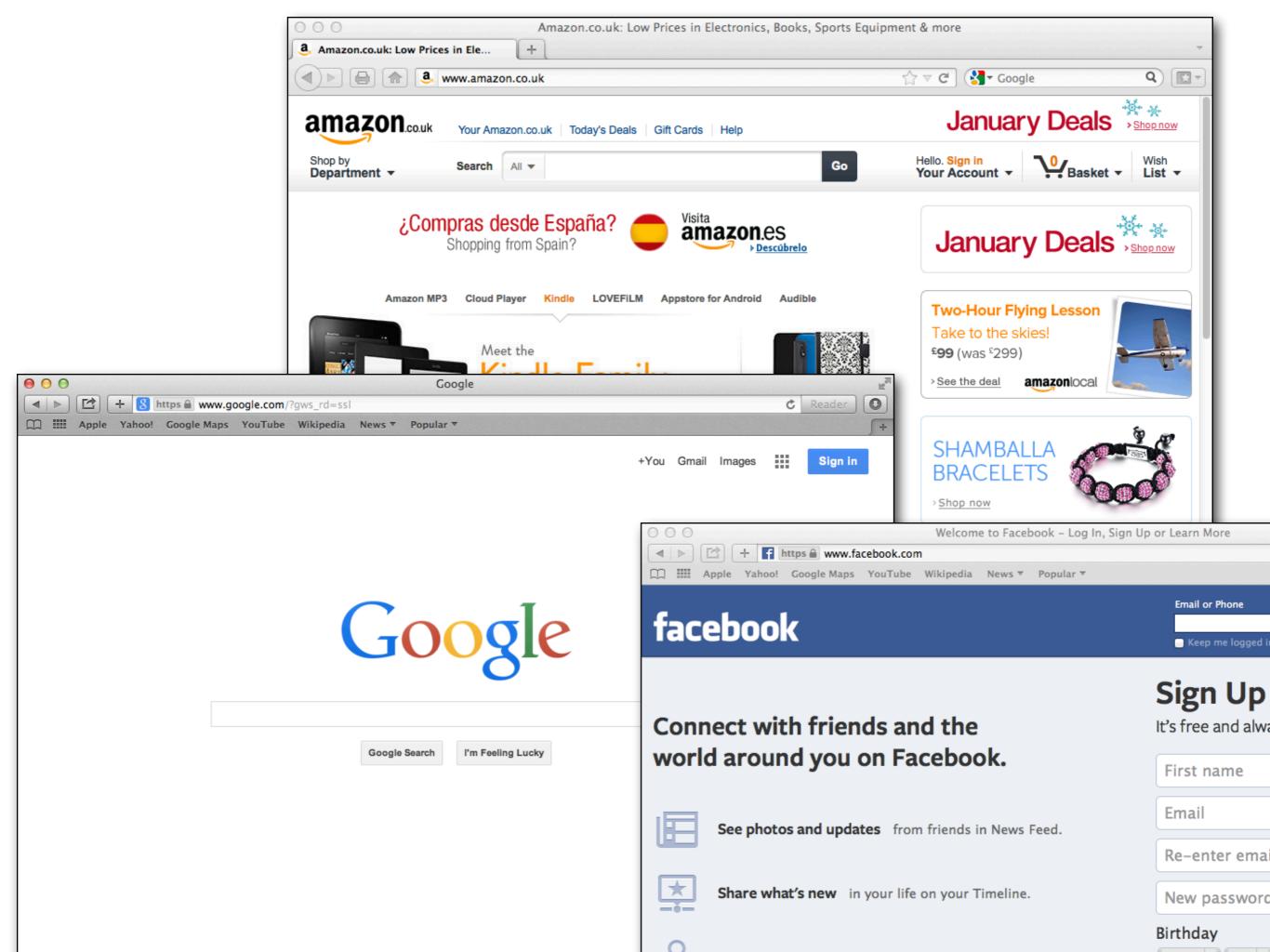
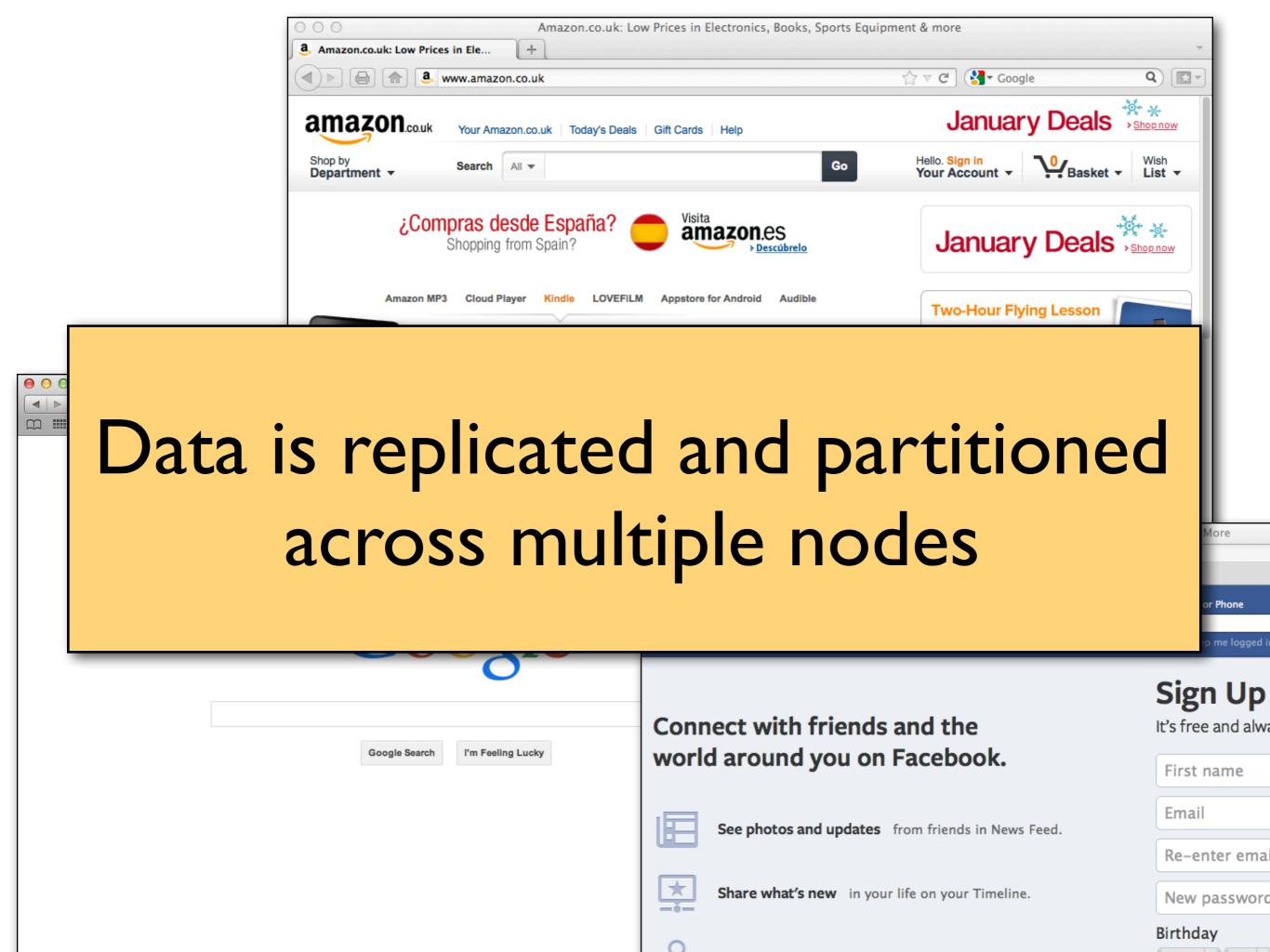
Reasoning about data consistency in distributed systems

Alexey Gotsman

IMDEA Software Institute, Madrid, Spain





Data centres across the world

Disaster-tolerance, minimising latency

Data centres across the world

Disaster-tolerance, minimising latency

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Disaster-tolerance, minimising latency

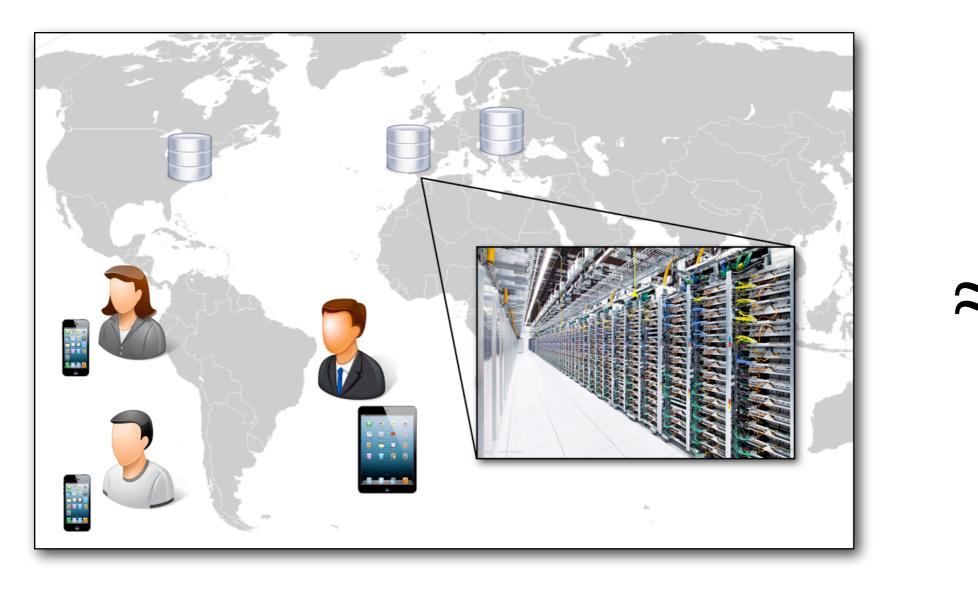
With thousands of machines inside

Load-balancing, fault-tolerance



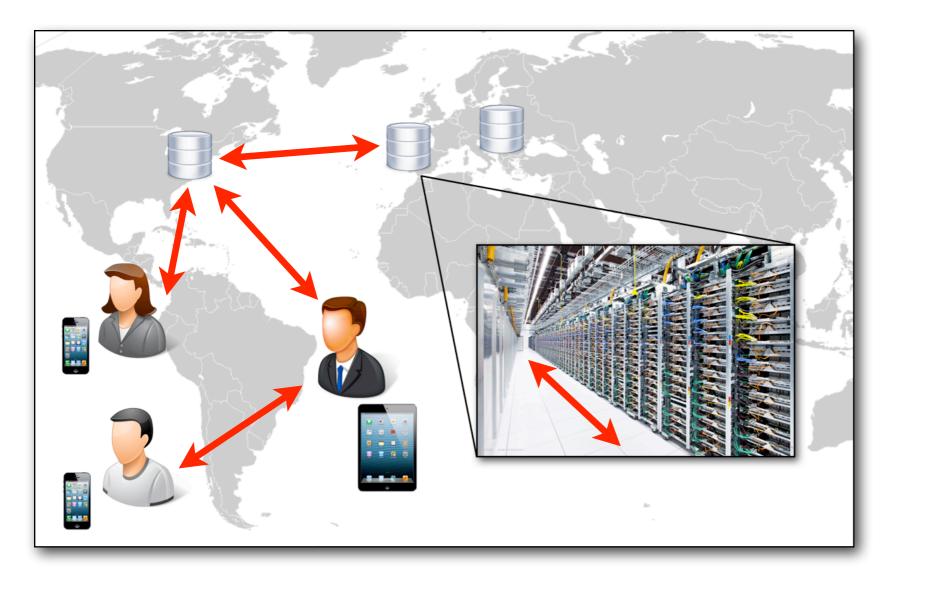






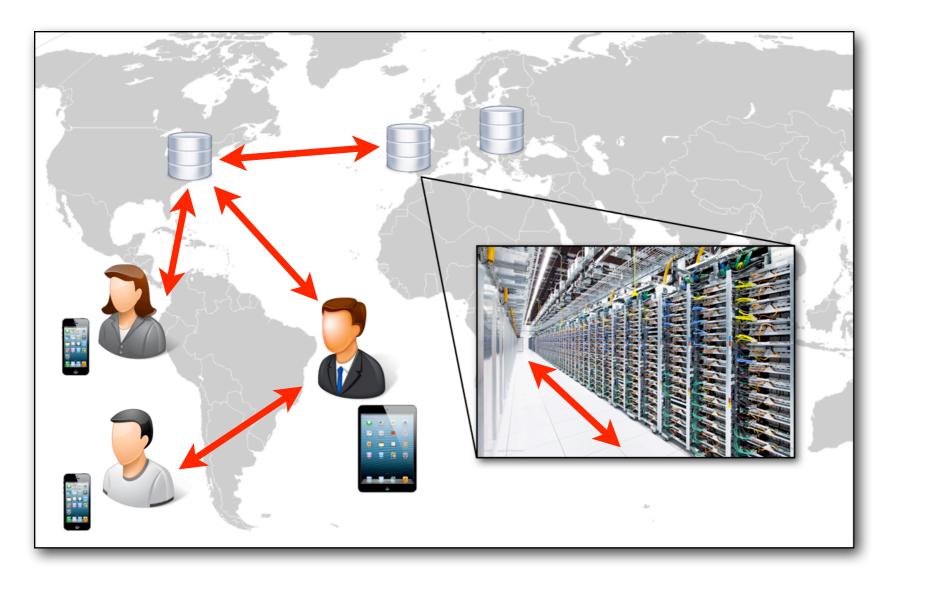


 Strong consistency model: the system behaves as if it processes requests serially on a centralised database - linearizability, serializability





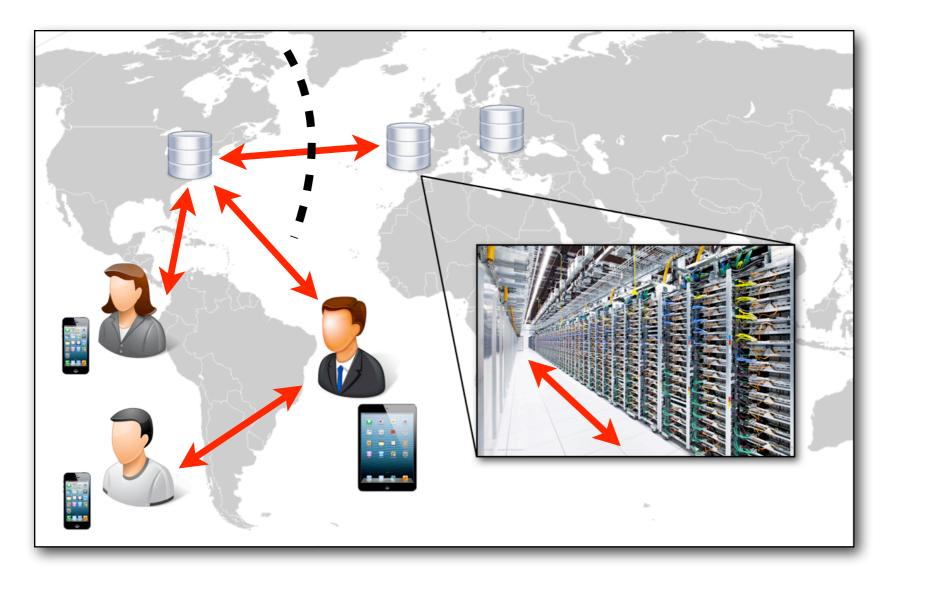
- Strong consistency model: the system behaves as if it processes requests serially on a centralised database - linearizability, serializability
- Requires synchronisation: contact other replicas when processing a request





 \sim

- Expensive: communication increases latency
- Impossible: either strong Consistency or Availability in the presence of network Partitions [CAP theorem]





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- Expensive: communication increases latency
- Impossible: either strong Consistency or Availability in the presence of network Partitions [CAP theorem]

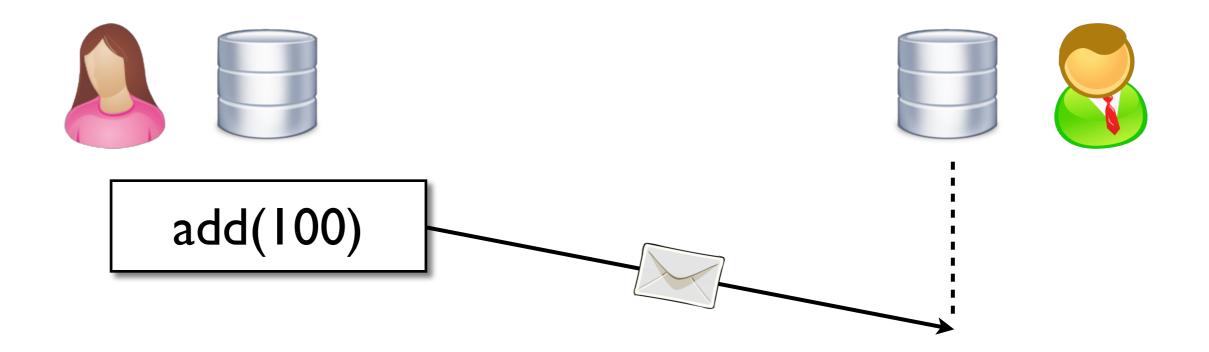




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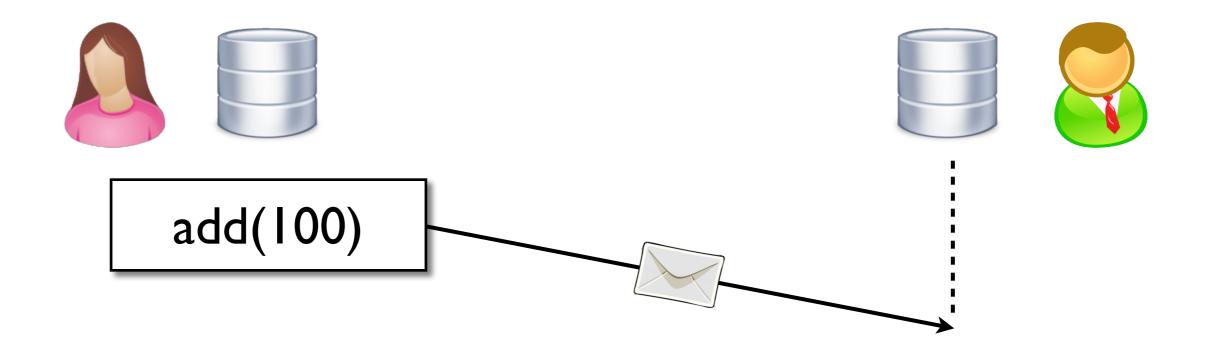
- Expensive: communication increases latency
- Impossible: either strong Consistency or Availability in the presence of network Partitions [CAP theorem]

Relaxing synchronisation



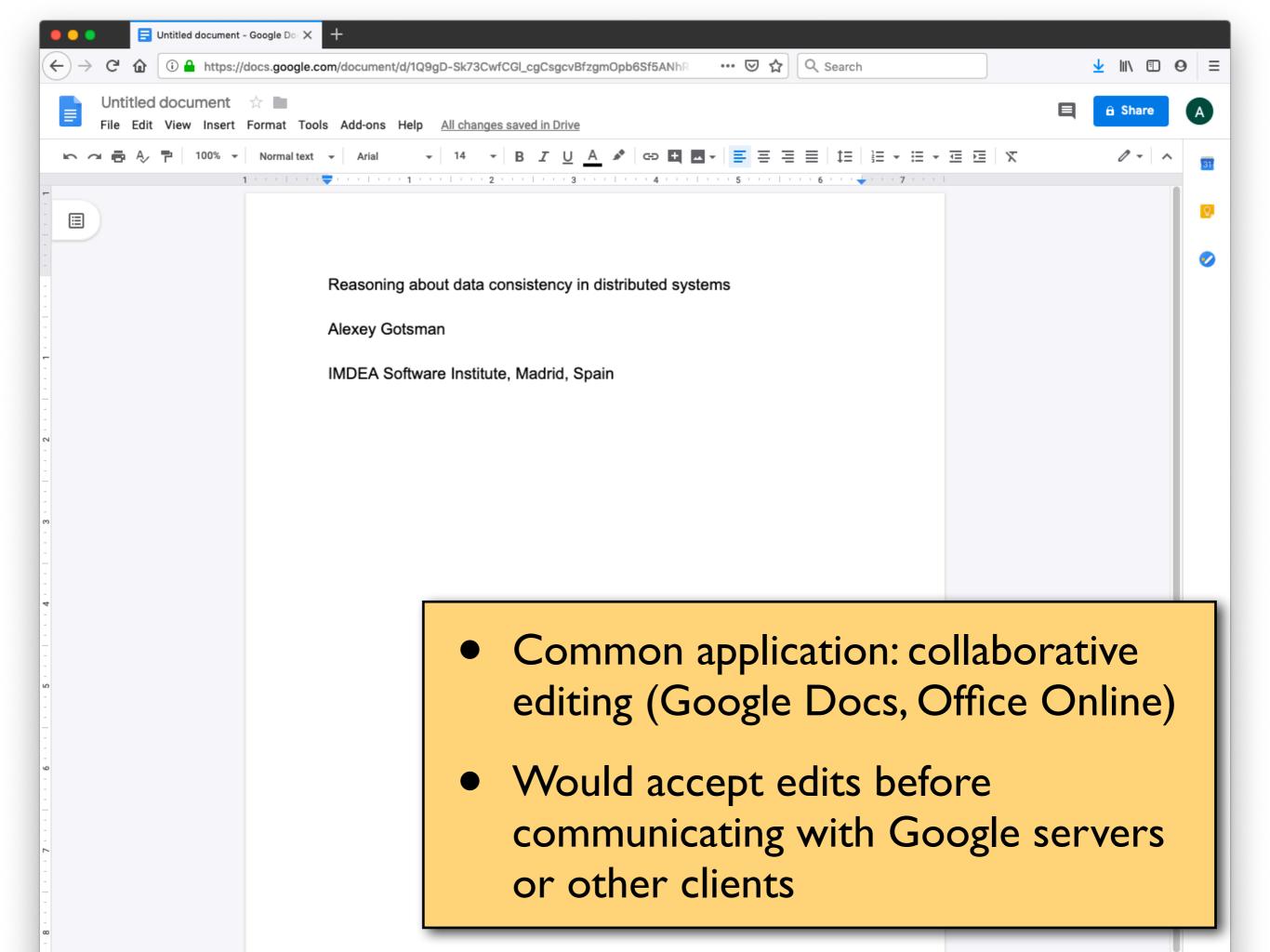
Process an update locally, propagate effects to other replicas later

Relaxing synchronisation



Process an update locally, propagate effects to other replicas later

- + Better scalability & availability
- Weakens consistency: deposit seen with a delay



NoSQL data stores

New generation of data stores with high scalability and low latency, but weak consistency







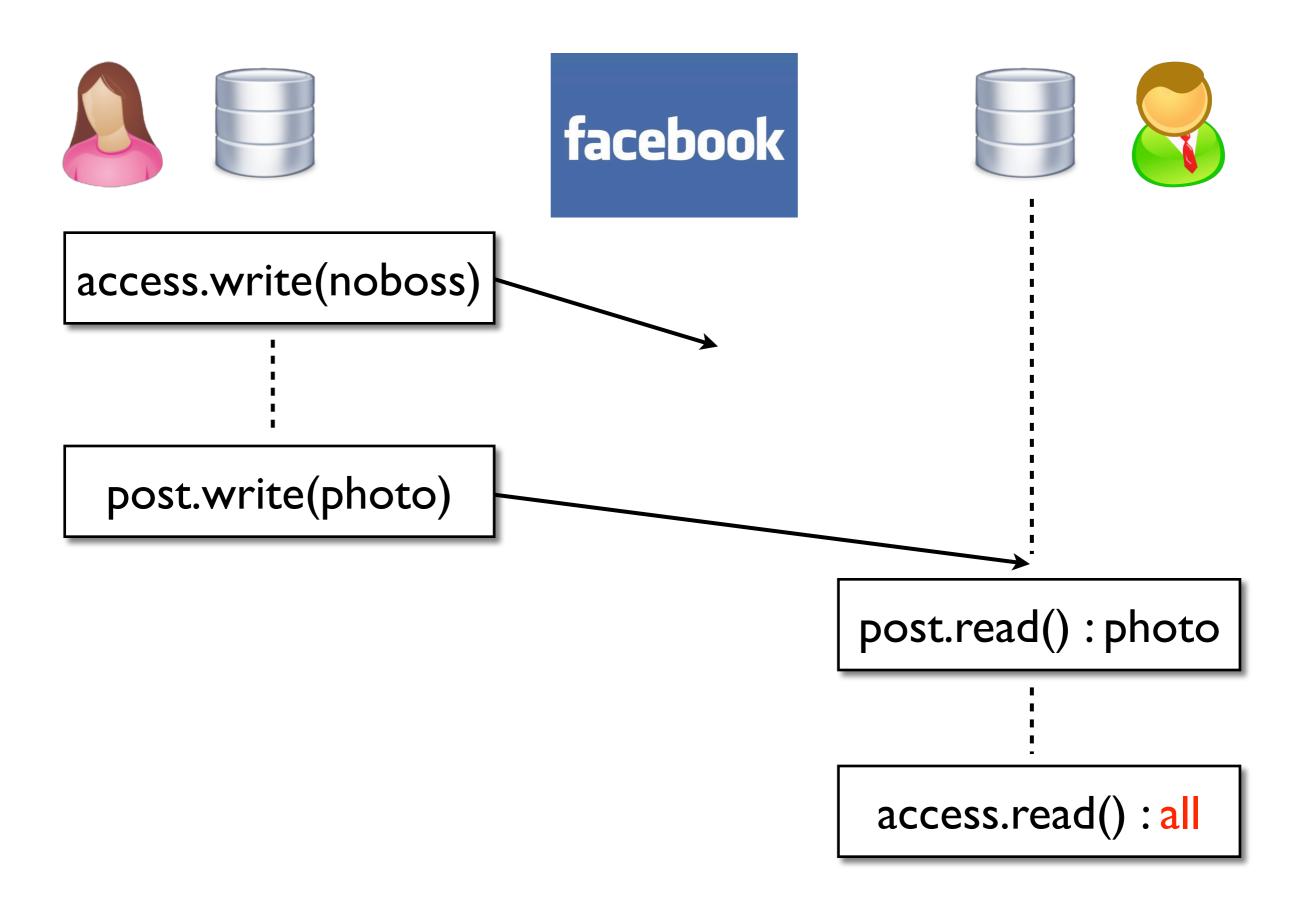


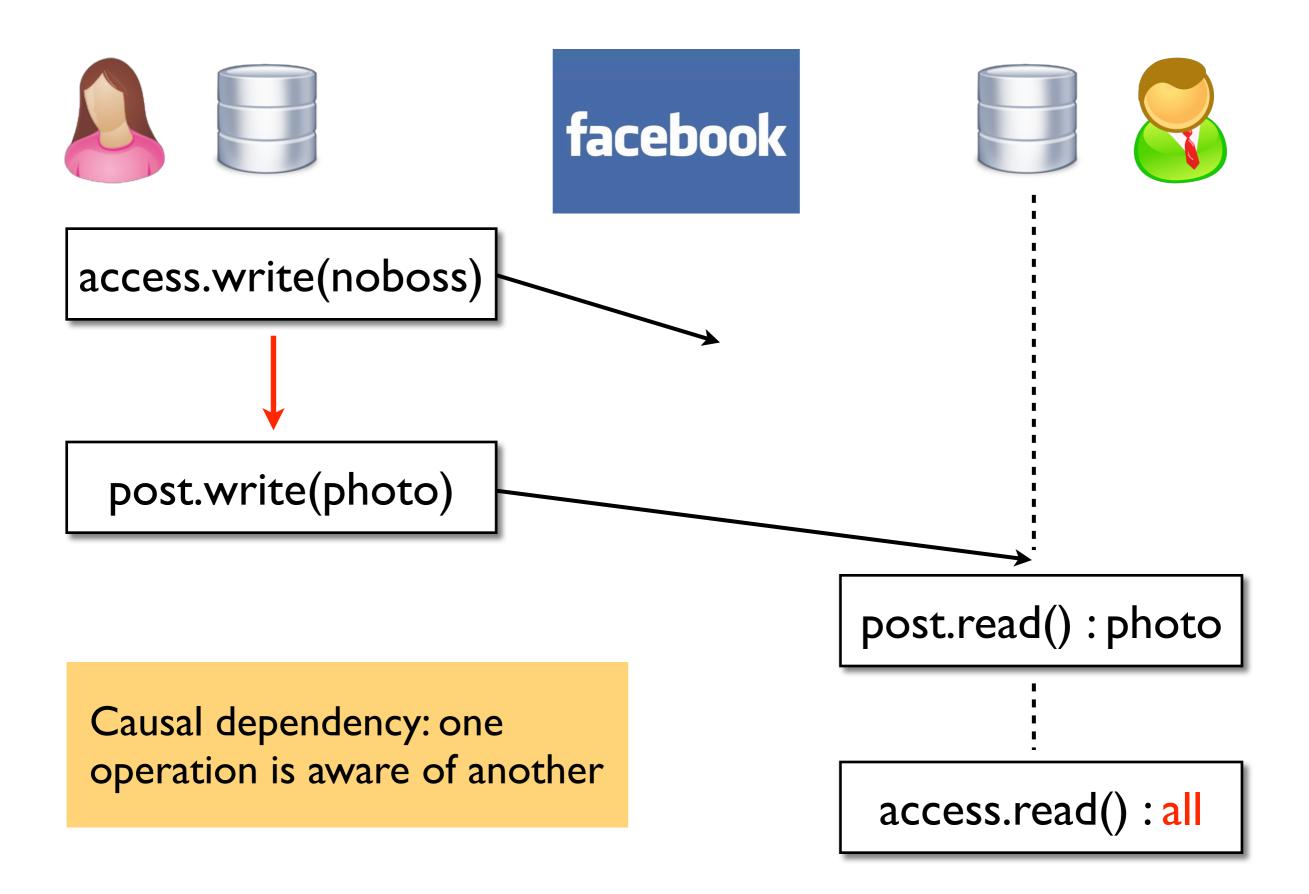
So what consistency guarantees do they provide?

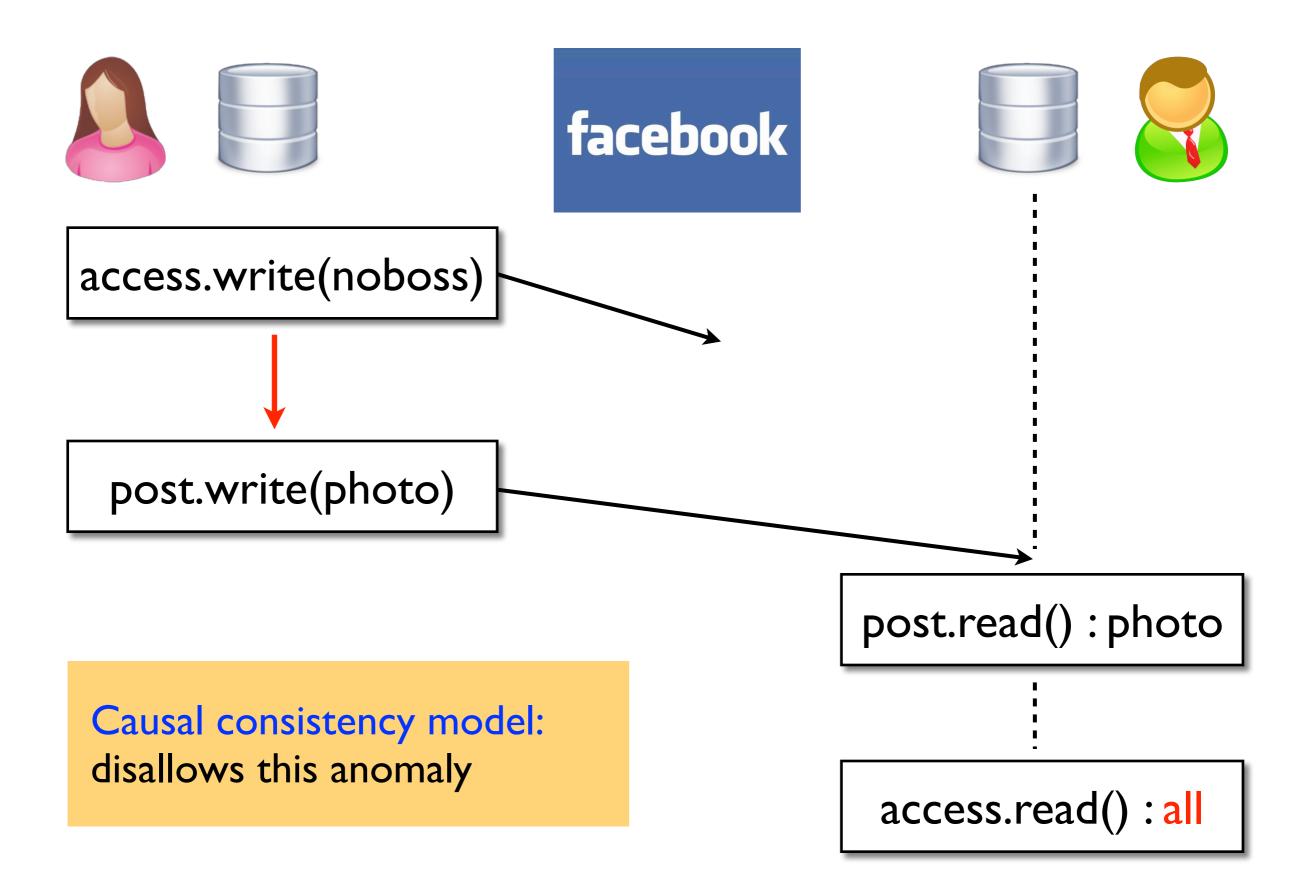


access.write(noboss)

post.write(photo)







Poor guidelines on how to use the weakly consistent data stores: are we weakening consistency too much, too little, just right?

Poor guidelines on how to use the weakly consistent data stores: are we weakening practice

> Building reliable distributed systems at a worldwide scale demands trade-offs between consistency and availability.

BY WERNER VOGELS

Eventually Consistent

AT THE FOUNDATION of Amazon's cloud computing are infrastructure services such as Amazon's S3 (Simple Storage Service), SimpleDB, and EC2 (Elastic Compute Cloud) that provide the resources for constructing Internet-scale computing platforms and a great variety of applications. The requirements placed on these infrastructure services are very strict; they need to score high marks in the areas of security, scalability, availability, performance, and cost-effectiveness, and they need to meet these requirements while serving millions of customers around the globe, continuously.

Under the covers these services are massive distributed systems that operate on a worldwide scale. This scale creates additional challenges, because when a system processes trillions and trillions of requests, events that normally have a low probability of occurrence are now guaranteed to happen and must be accounted for upfront in the design and architecture of the system. Given the worldwide scope of these systems, we use replication techniques ubiquitously to guarantee consistent performance and high availability. Although replication brings us closer to our goals, it cannot achieve them in a perfectly

transparent manner, under a number of conditions the customers of these services will be confronted with the consequences of using replication techniques inside the services.

One of the ways in which this manifests itself is in the type of data consistency that is provided, particularly when many widespread distributed systems provide an eventual consistency model in the context of data replication. When designing these largescale systems at Amazon, we use a set of guiding principles and abstractions related to large-scale data replication and focus on the trade-offs between high availability and data consistency. Here, I present some of the relevant background that has informed our approach to delivering reliable distributed systems that must operate on a global scale. (An earlier version of this article appeared as a posting on the "All Things Distributed" Weblog and was greatly improved with the help of its readers.)

Historical Perspective

In an ideal world there would be only one consistency model: when an update is made all observers would see that update. The first time this surfaced as difficult to achieve was in the database systems of the late 1970s. The best "period piece" on this topic is "Notes on Distributed Databases" by Bruce Lindsay et al.5 It lays out the fundamental principles for database replication and discusses a number of techniques that deal with achieving consistency. Many of these techniques try to achieve distribution transparency-that is, to the user of the system it appears as if there is only one system instead of a number of collaborating systems. Many systems during this time took the approach that it was better to fail the complete system than to break this transparency.2

In the mid-1990s, with the rise of larger Internet systems, these practices were revisited. At that time people began to consider the idea that availability was perhaps the most impor-

"If no new updates are made to the database, then replicas will eventually reach a consistent state"

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2008

TOWARDS A CLOUD COMPUTING RESEARCH AGENDA

Ken Birman, Gregory Chockler, Robbert van Renesse

This particular example is a good one because, as we'll see shortly, if there was a single overarching theme within the keynote talks, it turns out to be that strong synchronization of the sort provided by a locking service must be avoided like the plague. This doesn't diminish the need for a tool like Chubby; when locking actually can't be avoided, one wants a reliable, standard, provably correct

2008

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F1: A Distributed SQL Database That Scales 2013

Jeff Shute Chad Whipkey David Menestrina Radek Vingralek Eric Rollins Stephan Ellner Traian Stancescu Bart Samwel Mircea Oancea John Cieslewicz Himani Apte

Ben Handy Kyle Littlefield Ian Rae*

Google, Inc. *University of Wisconsin-Madison

ABSTRACT

F1 is a distributed relational database system built at Google to support the AdWords business. F1 is a hybrid database that combines high availability, the scalability of NoSQL systems like Bigtable, and the consistency and us-

consistent and correct data.

Designing applications to cope with concurrency anomalies in their data is very error-prone, timeconsuming, and ultimately not worth the performance gains.

Strong vs weak consistency

- Pay-off from weakening consistency often worth it: higher scalability, lower latency in geo-distribution, offline access
 - Both strong and weak systems used in industry
- But programmers need help in using it:
 - Programming abstractions for weak consistency
 - Methods for reasoning about how weakening consistency affects application correctness

Also centralised SQL databases

Don't provide strong consistency either by default or at all: to exploit single-node concurrency



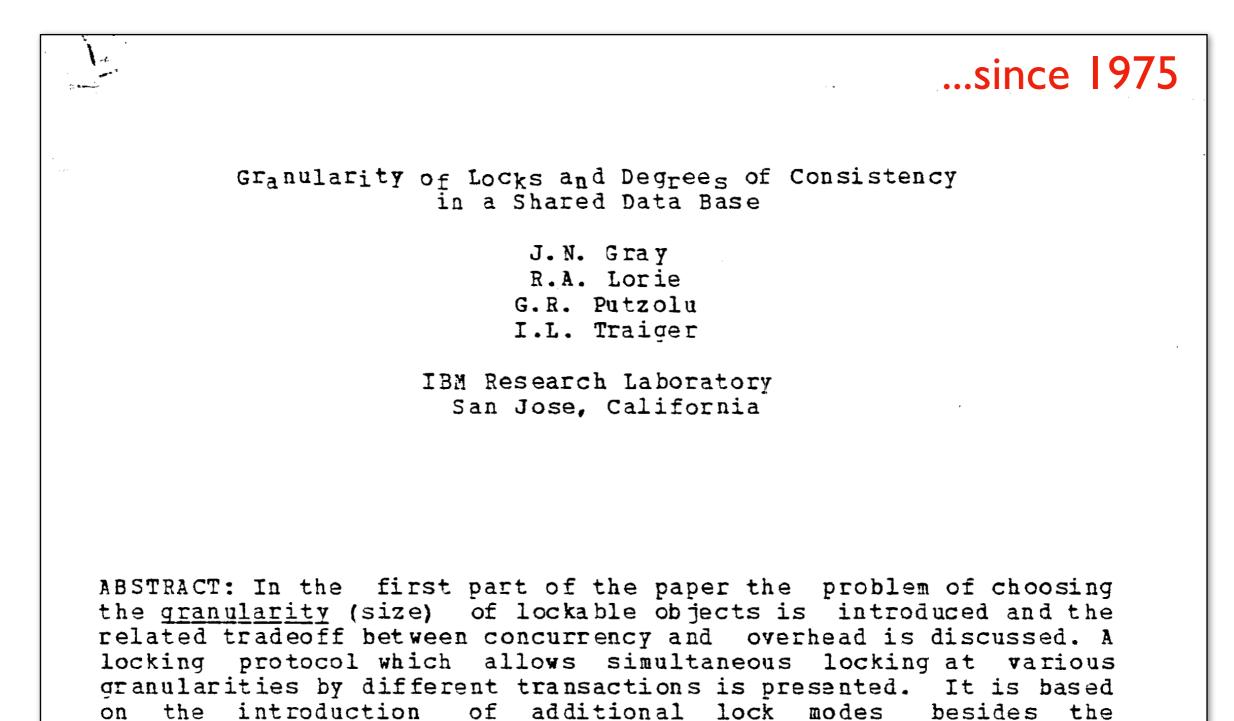






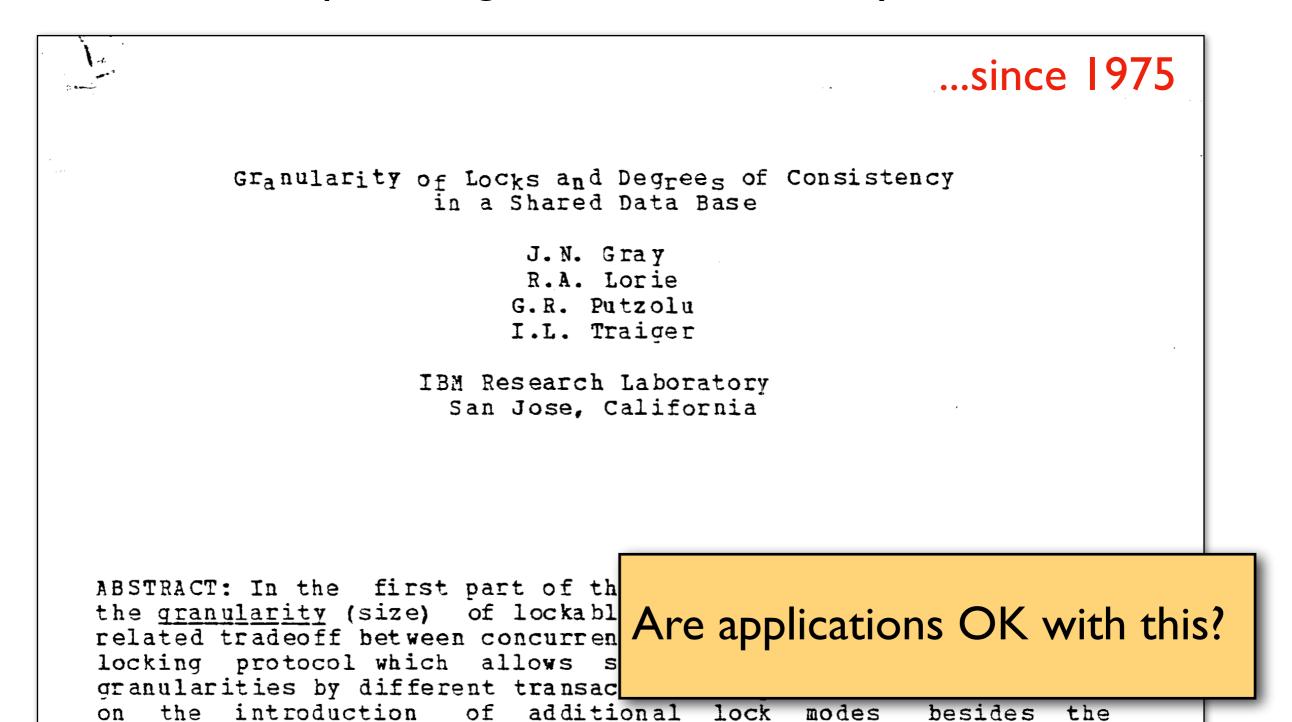
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Don't provide strong consistency either by default or at all: to exploit single-node concurrency



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[SIGMOD'17]

ACIDRain: Concurrency-Related Attacks on Database-Backed Web Applications

Todd Warszawski, Peter Bailis Stanford InfoLab

ABSTRACT

In theory, database transactions protect application data from corruption and integrity violations. In practice, database transactions frequently execute under weak isolation that exposes programs to a range of concurrency anomalies, and programmers may fail to correctly employ transactions. While low transaction volumes mask many potential concurrency-related errors under normal operation, determined adversaries can exploit them programmatically for fun and profit. In this paper, we formalize a new kind of attack on database-backed applications called an ACIDRain attack, in which an adversary systematically exploits concurrency-related vulnerabilities via programmatically accessible APIs. These attacks are not theoretical: ACIDRain attacks have already occurred in a handful of applications in the wild, including one attack which bankrupted a popular Bitcoin exchange. To proactively detect the potential for ACIDRain attacks, we extend the theory of weak isolation to analyze latent potential for non-serializable behavior under concurrent web API calls. We introduce a language-agnostic method for detecting potential isolation anomalies in web applications, called Abstract Anomaly Detection (2AD), that uses dynamic traces of database accesses to efficiently reason about the space of possible concurrent interleavings. We apply a prototype 2AD analysis tool to 12 popular self-hosted eCommerce applications written in four languages and deployed on over 2M websites. We identify and verify 22 critical ACIDRain attacks that allow attackers to corrupt store inventory, over-spend gift cards, and steal inventory.

1	def withdraw(amt, user_id):	(a)
2	$bal = readBalance(user_id)$	
3	if (bal $\geq = amt$):	
4	writeBalance(bal – amt, user_id)	
	· · · · · · · · · · · · · · · · · · ·	

Figure 1: (a) A simplified example of code that is vulnerable to an ACIDRain attack allowing overdraft under concurrent access. Two concurrent instances of the withdraw function could both read balance 100, check that $100 \ge 99$, and each allow 99 to be withdrawn, resulting 198 total withdrawals. (b) Example of how transactions could be inserted to address this error. However, even this code is vulnerable to attack at isolation levels at or below Read Committed, unless explicit locking such as SELECT FOR UPDATE is used. While this scenario closely resembles textbook examples of improper transaction use, in this paper, we show that widely-deployed eCommerce applications are similarly vulnerable to such ACIDRain attacks, allowing corruption of application state and theft of assets.

[SIGMOD'17]

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5 writeBalance(bal – amt, user_id)

6 commit()

No! E-commerce applications can be hacked by exploiting weak consistency of back-end databases

corruption of application state and theft of assets.

Weak shared-memory models

• Multicore processors: x86, ARM



Multiprocessor ~ distributed system

Programming languages: C/C++, Java
 Due to compiler optimisations



This course

- Programming abstractions for weak consistency
- Methods for specification
- Methods and tools for reasoning about application correctness and consistency needs
- Implementing strong consistency

Strong consistency and the CAP theorem

Data model

- Database system manages a set of objects:
 Obj = {x, y, z...}
- Objects associated with types Type = $\{\tau, ...\}$
- For each type $\tau \in Type$:
 - ► Set of operations Op_T, including arguments
 - ► Return values:Val_T

Data model

- Integer register
 - Opintreg = {read, write(k) | $k \in \mathbb{Z}$ }
 - ► $Val_{intreg} = \mathbb{Z} \cup \{ok\}$
- Counter:
 - Opcounter = {read, add(k) | $k \in \mathbb{N}$ }
 - $Val_{counter} = \mathbb{N} \cup \{ok\}$

Sequential semantics

- Semantics in an ordinary programming language
- For each type $\tau \in Type$: set of states State_{τ}, initial state $\sigma_0 \in State_{\tau}$
 - State_{intreg} = \mathbb{Z}
 - State_{counter} = \mathbb{N}
- Semantics of an operation op:
 - $[op]_{val} \in State_{\tau} \rightarrow Value_{\tau}$
 - ▶ $[op]_{state} \in State_{T} \rightarrow State_{T}$

Register semantics

- State = \mathbb{Z}
- $[[write(k)]]_{state}(\sigma) = k$
- $\llbracket write \rrbracket_{val}(\sigma) = ok$
- $[[read]]_{state}(\sigma) = \sigma$
- $\llbracket read \rrbracket_{val}(\sigma) = \sigma$

Counter semantics

- State = \mathbb{N}
- $[add(k)]_{state}(\sigma) = \sigma + k$
- $[add(k)]_{val}(\sigma) = ok$
- $[[read]]_{state}(\sigma) = \sigma$
- $[[read]]_{val}(\sigma) = \sigma$

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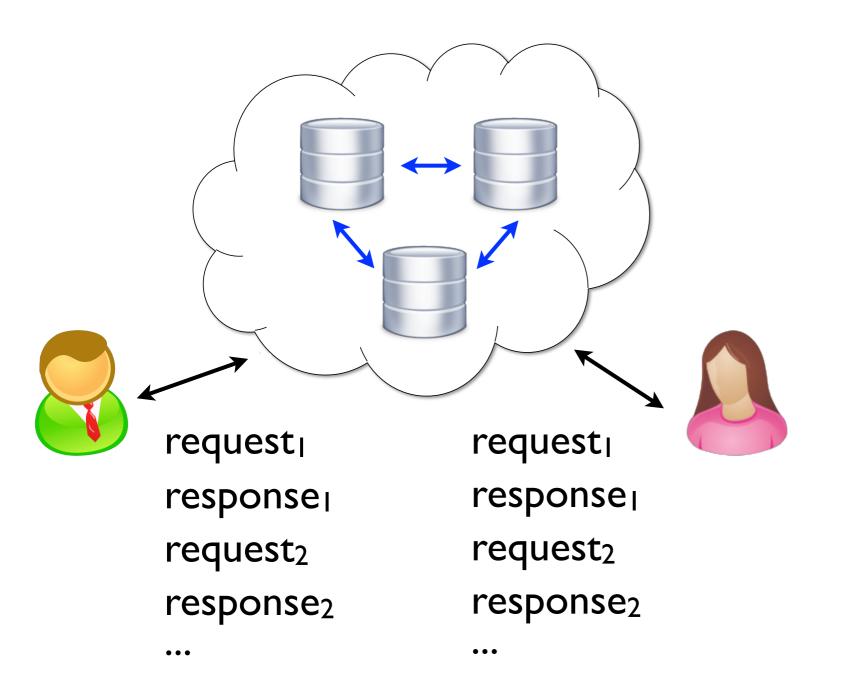
read-only operation: $[op]_{state}(\sigma) = \sigma$

Counter semantics

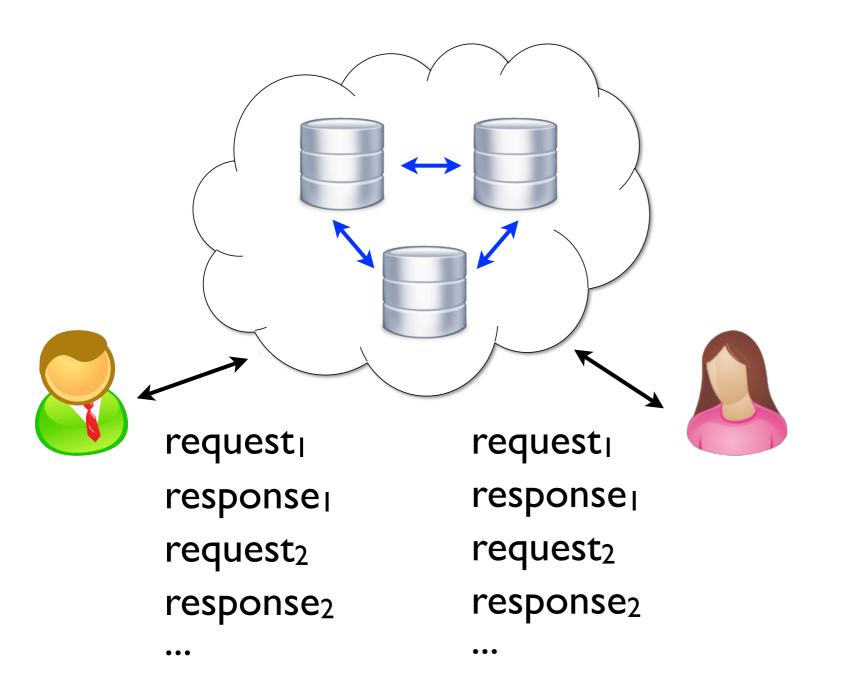
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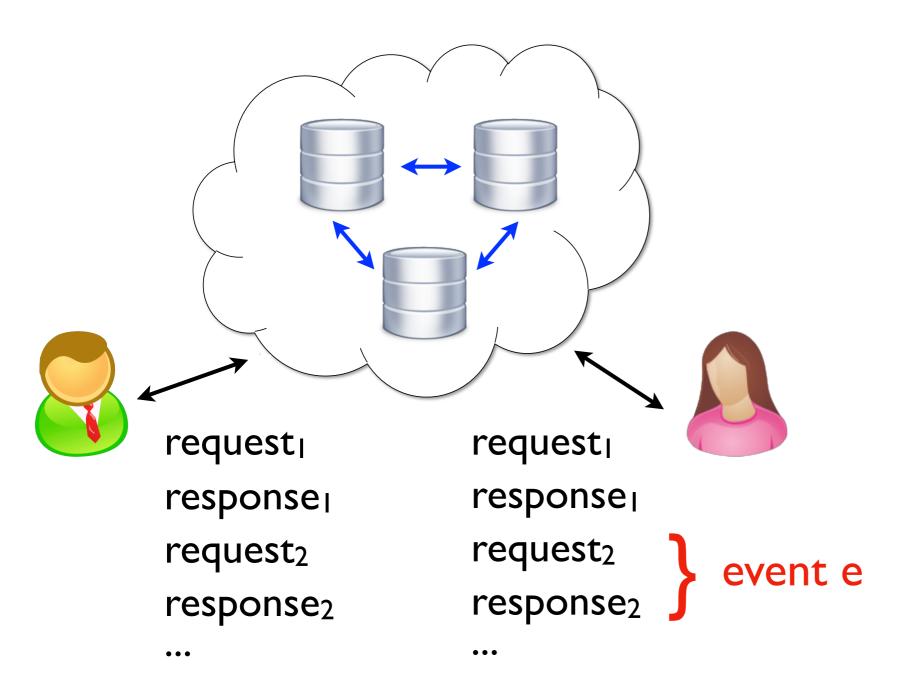
update operation



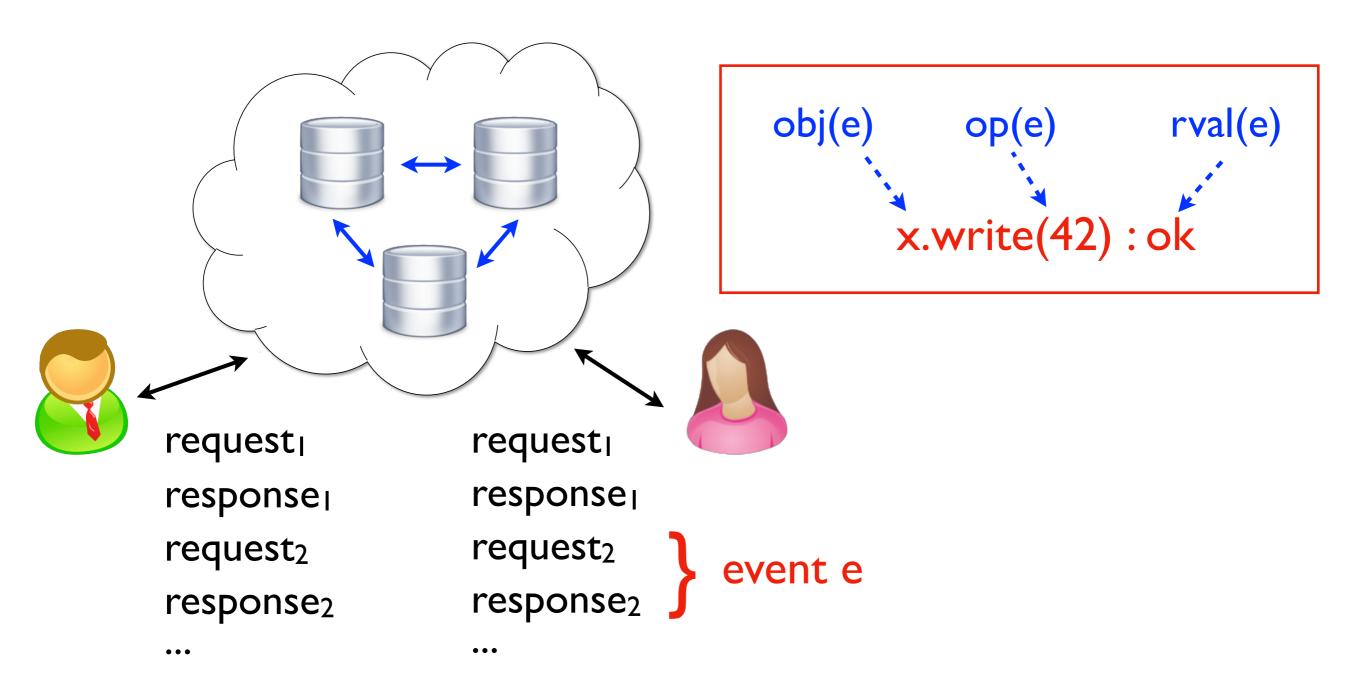
Clients issue requests and get responses: history records the interactions in a single execution



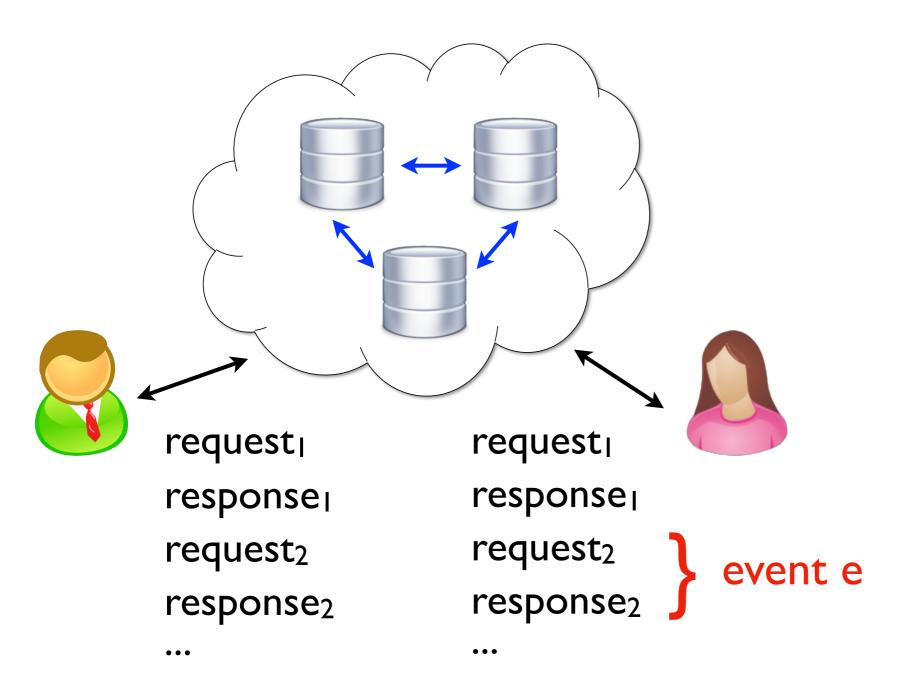
Assume every request yields a response No next request until the previous one responded

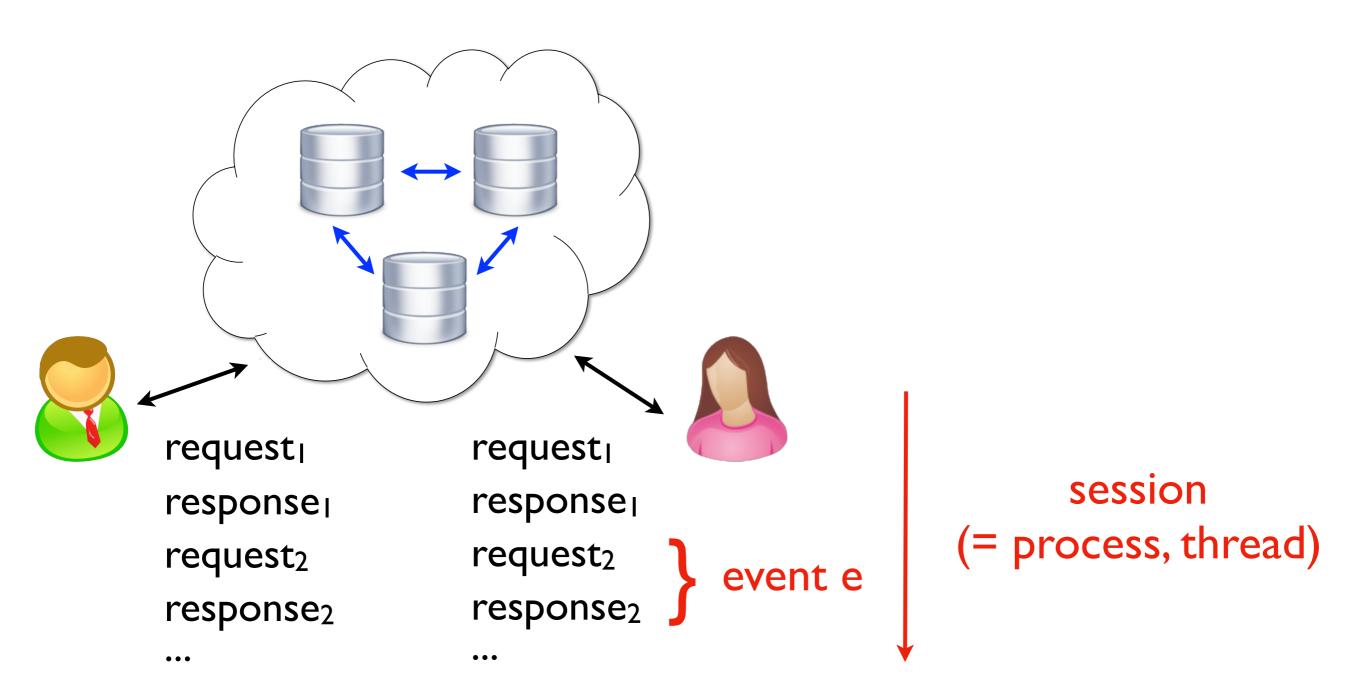


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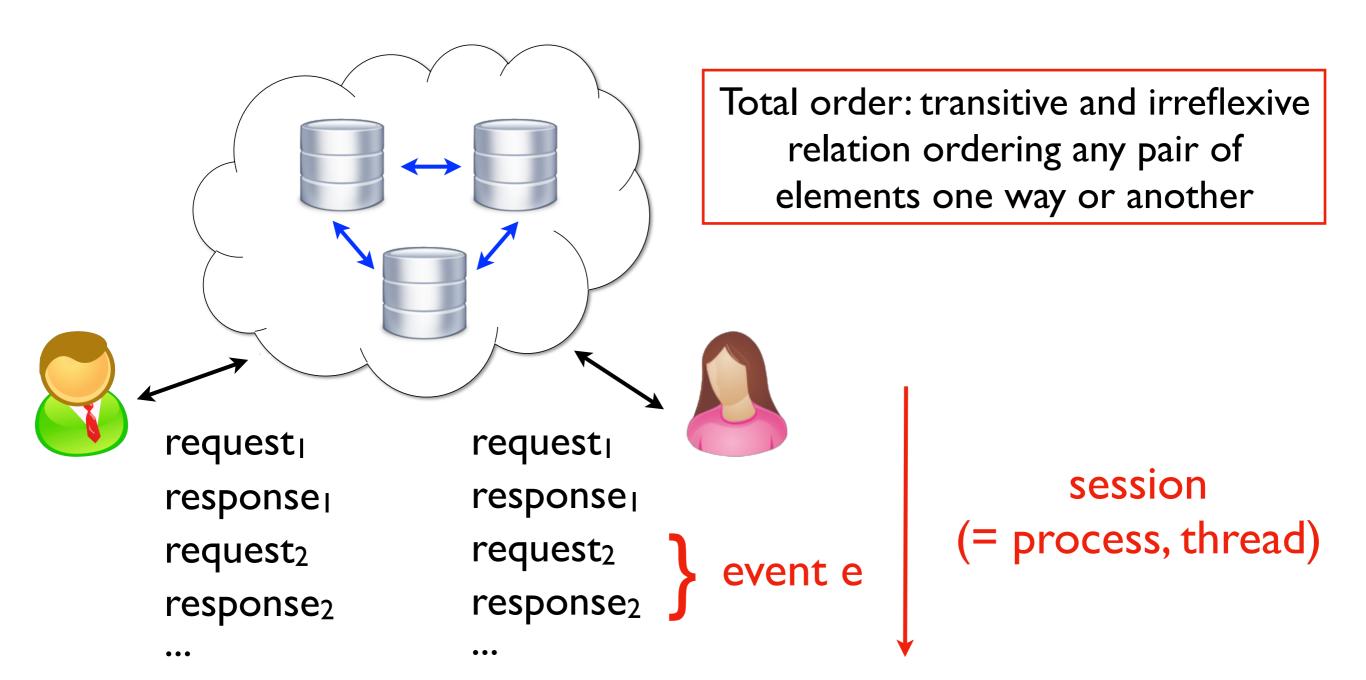


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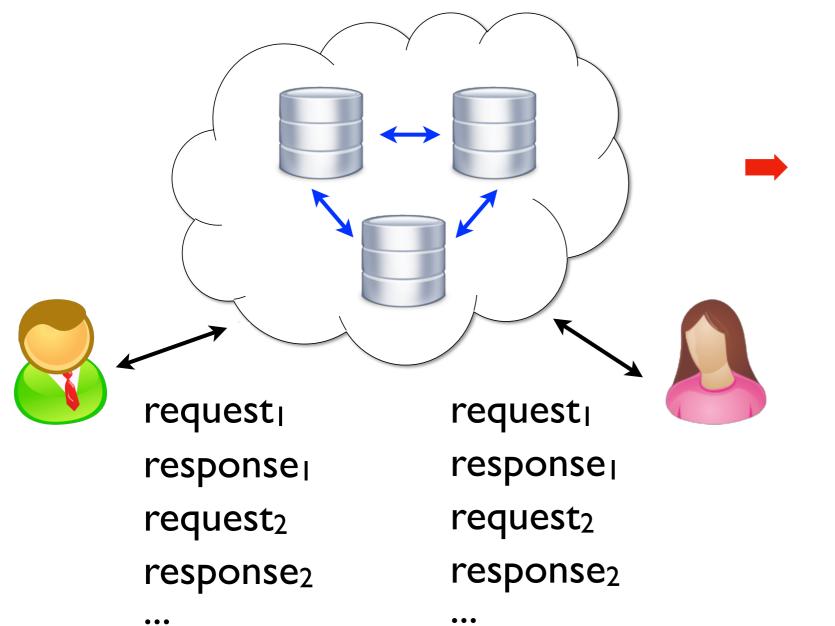




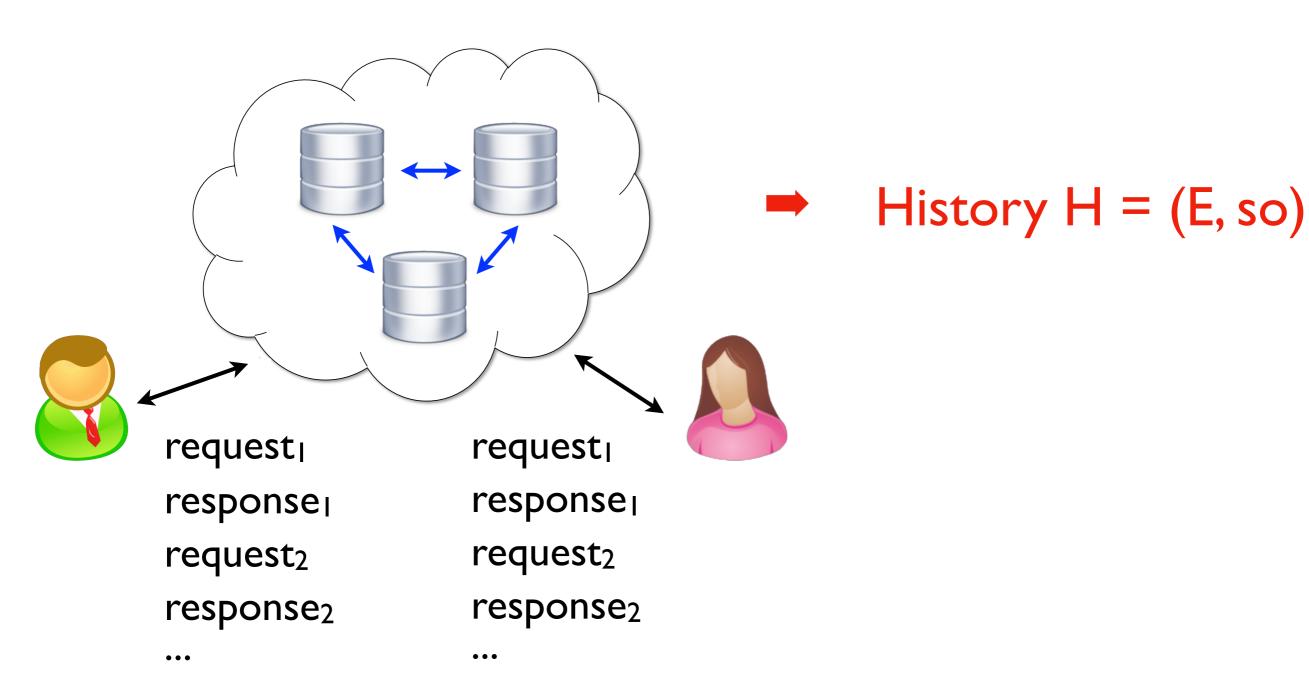
Session order so: the order in which events are issued: union of total per-client total orders



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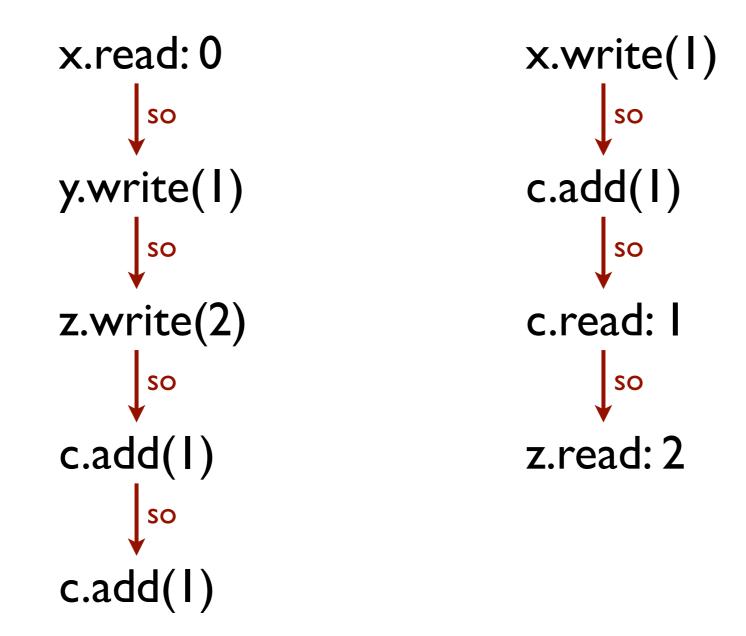


History H = (E, so)

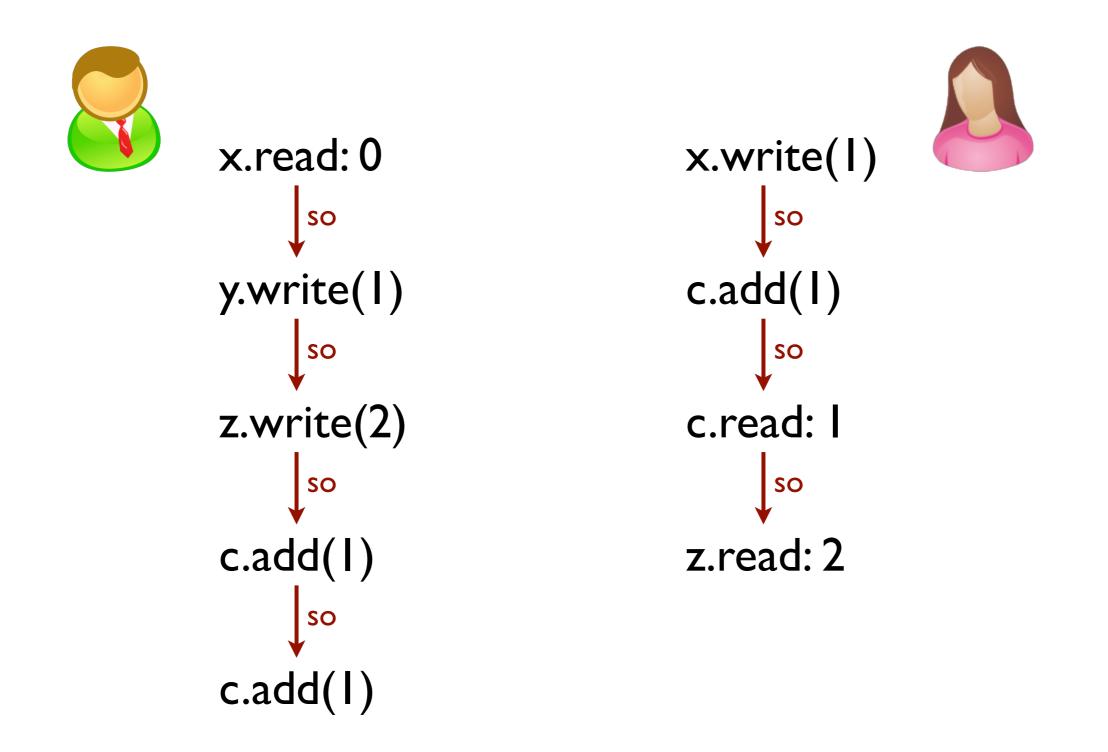


Consistency model - a set of histories \mathscr{H} : the set of allowed database behaviours

Visualising histories



Visualising histories



• Consistency model *H*: behaviour of the database under arbitrary clients

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- Program P → set of all executions [[P]] under arbitrary behaviour of the database

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```
P:
rl = x.read();
r2 = x.read();
y.write(rl==r2);
```

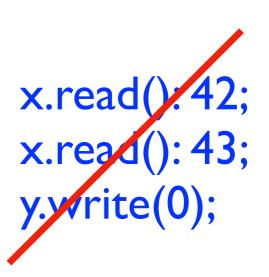
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```
P: [[P]]:
rl = x.read(); x.read():42; x.read():42;
r2 = x.read(); x.read():42; x.read():43;
y.write(rl==r2); y.write(l); y.write(0);
```

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P: rI = x.read();r2 = x.read();y.write(rI = r2);

[P]: x.read(): 42; x.read(): 42; x.read(): 42; x.read(): 43; y.write(1); y.write(0);



[P, *H***]**:

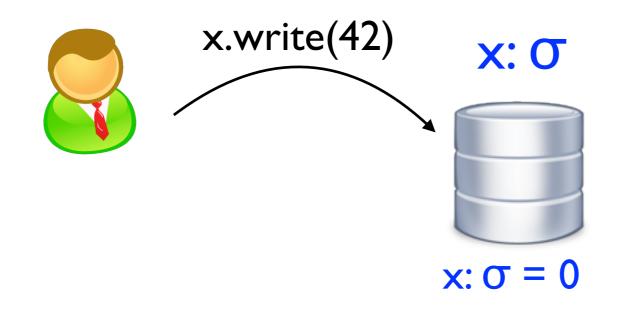
Defining a consistency model

- Operational specification: by an idealised implementation
- Axiomatic specification: more declarative

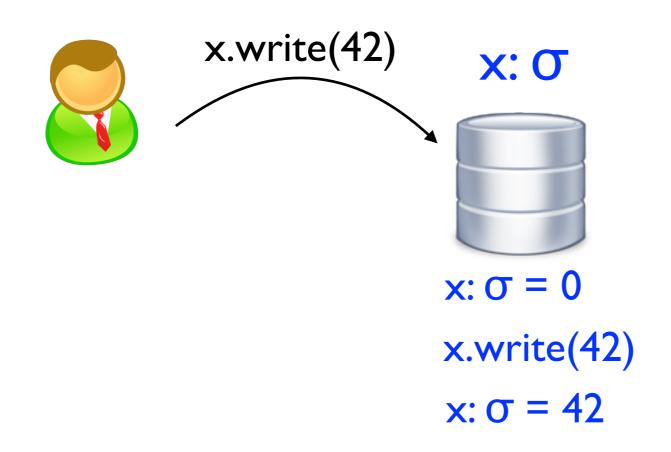
x: σ



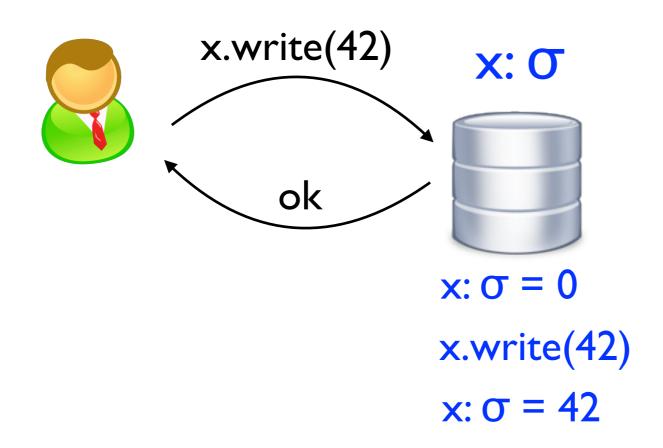
- Server with a single copy of all objects
- Clients send request to the server and wait for a reply
- Server processes operation sequentually in the receipt order



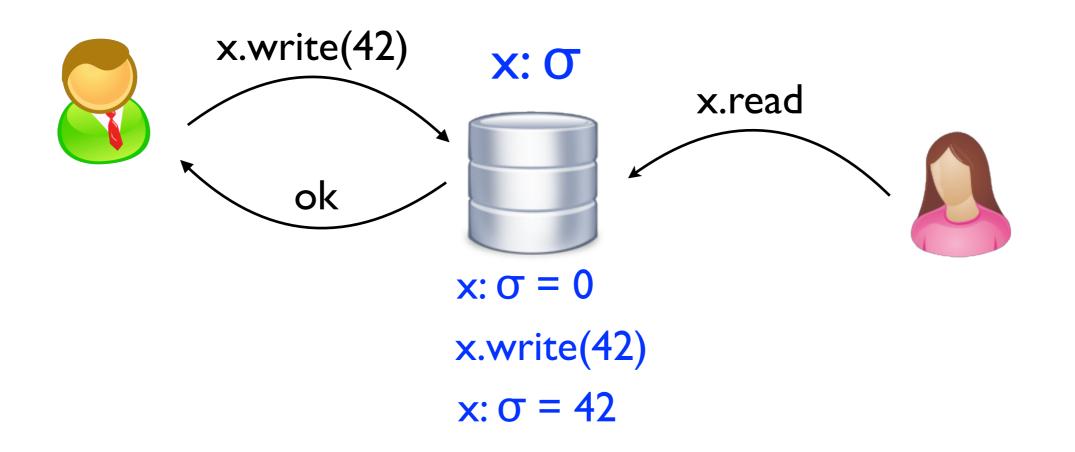
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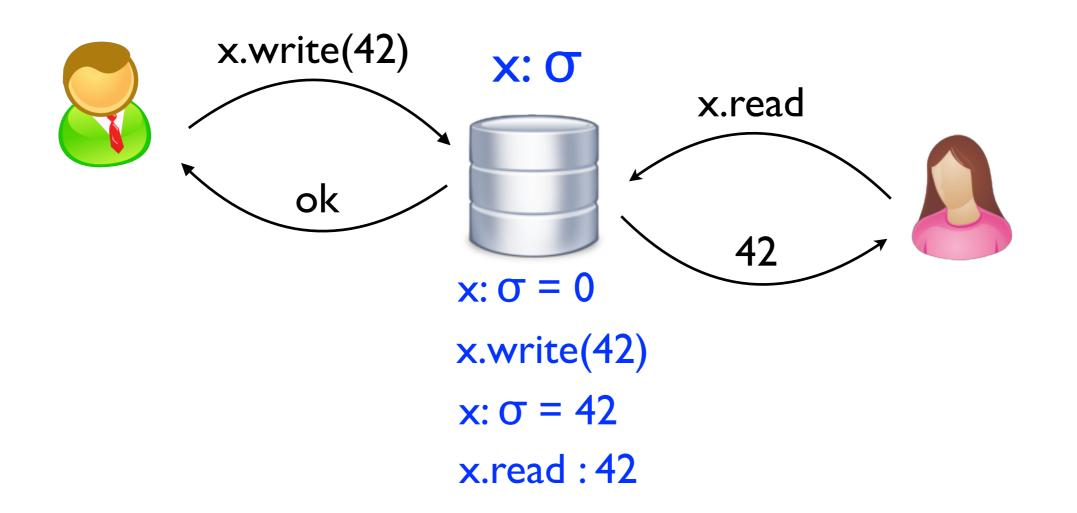
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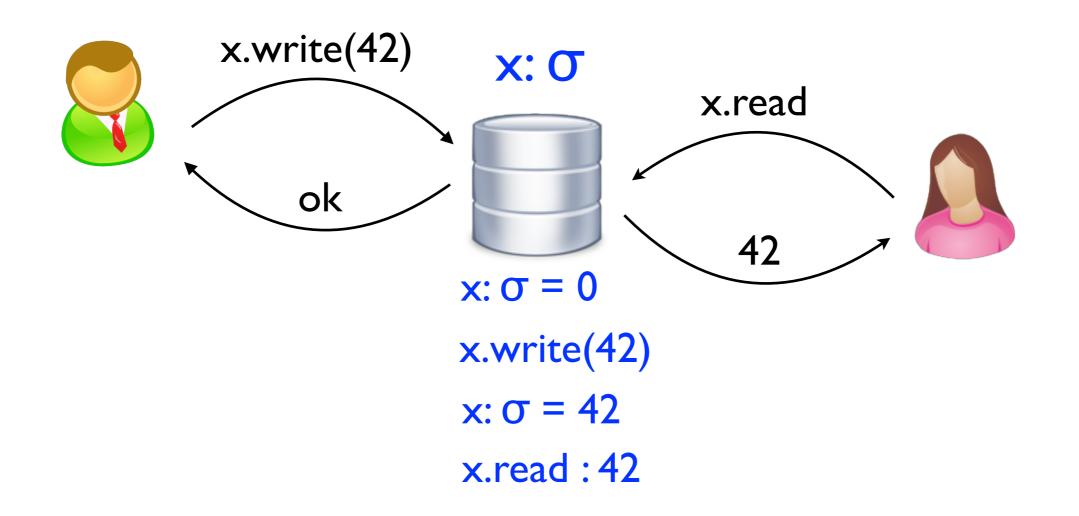
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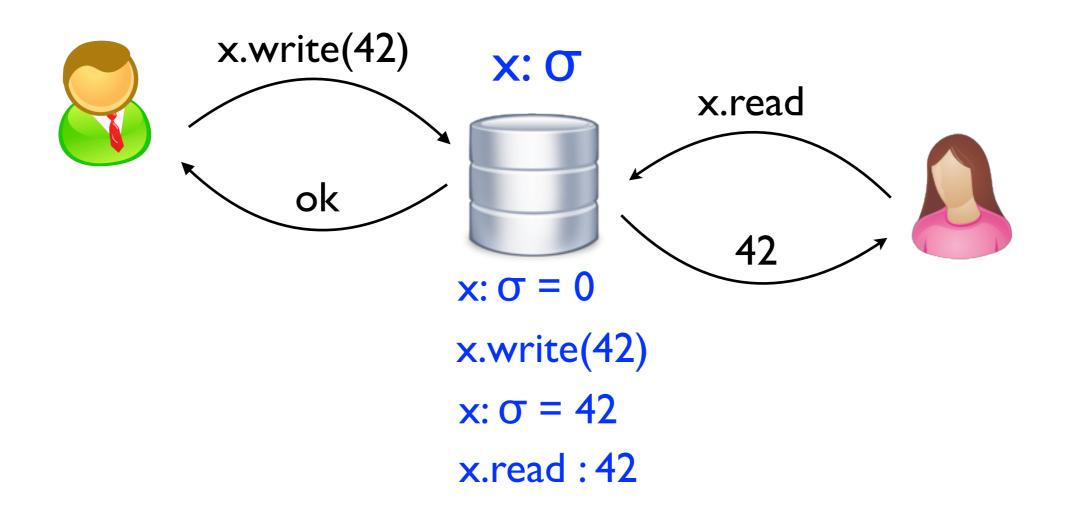
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- Server processes operation sequentually in the receipt order



Could write a formal operational semantics: maintain the state of the database, clients and sets of messages between them



- Consistency model = {H | ∃ execution with history H produced by the abstract implementation}
- Sequential consistency: one form of strong consistency
- Weaker than linearizability: takes into acount the duration of operations

Operational specifications

- Let one understand intuitions behind implementations
- May become unwieldy for weaker consistency models
- Sometimes overspecify behaviour

Axiomatic specifications

- Choose a set of relations over events: r₁, ..., r_n
 Abstractly specify essential information about how operations are processed inside the system
- Abstract execution $(H, r_1, ..., r_n) = (E, so, r_1, ..., r_n)$
- Choose a set of axioms *A* constraining abstract executions
- Consistency model = {H | $\exists r_1, ..., r_n$. (H, $r_1, ..., r_n$) $\vDash \mathscr{A}$ }

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Axiomatic specifications

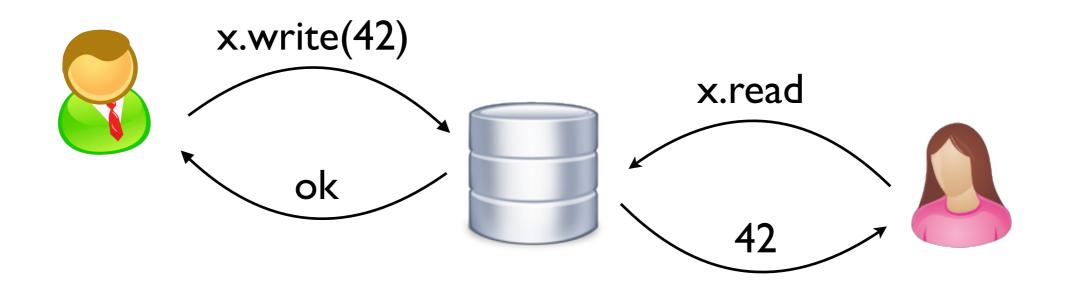
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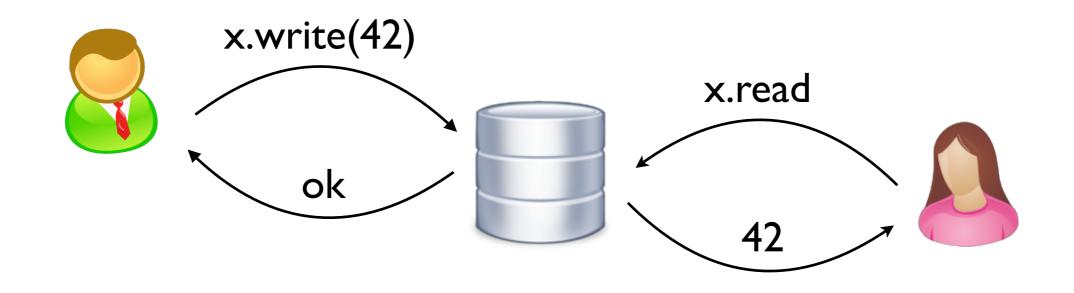
Sequential consistency axiomatically

An SC history can be explained by a total order over all events: the order in which the server processes client operations



Sequential consistency axiomatically

An SC history can be explained by a total order over all events: the order in which the server processes client operations



Abstract execution: (H, to) = (E, so, to), where $to \subseteq E \times E$

SC = {(E, so) | \exists total order to. (E, so, to) $\vDash \mathscr{A}_{SC}$ }

- I. so \subseteq to
- 2. The return value of each operation in E is computed from a state obtained by executing all operations on the same object preceding it in to

I. so \subseteq to

2. The return value of each operation in E is computed from a state obtained by executing all operations on the same object preceding it in to

 $\forall e \in E. type(obj(e)) = (\sigma_0, \llbracket - \rrbracket_{val}, \llbracket - \rrbracket_{state})$

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 $\forall e \in E. type(obj(e)) = (\sigma_0, \llbracket - \rrbracket_{val}, \llbracket - \rrbracket_{state})$

 $rval(e) = [op(e)]_{val}(\sigma)$

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$$\forall e \in E. type(obj(e)) = (\sigma_0, \llbracket - \rrbracket_{val}, \llbracket - \rrbracket_{state})$$

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$$\sigma = \llbracket op(e_n) \rrbracket_{state}(...\llbracket op(e_1) \rrbracket_{state}(\sigma_0))$$

 $e_1, ..., e_n = to^{-1}(e).select(obj(e)).sort(to)$

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Integer registers: a read returns the value written by the last preceding event in to (or 0 if there are none) x.write(0); x.write(42); x.read: 42

$$SC = \{(E, so) \mid \exists to. (E, so, to) \vDash \mathscr{A}_{SC}\}$$

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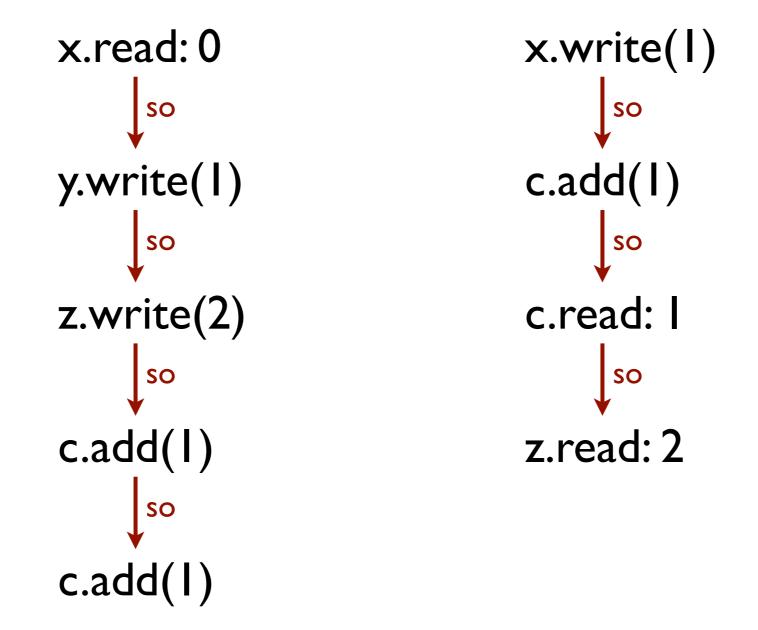
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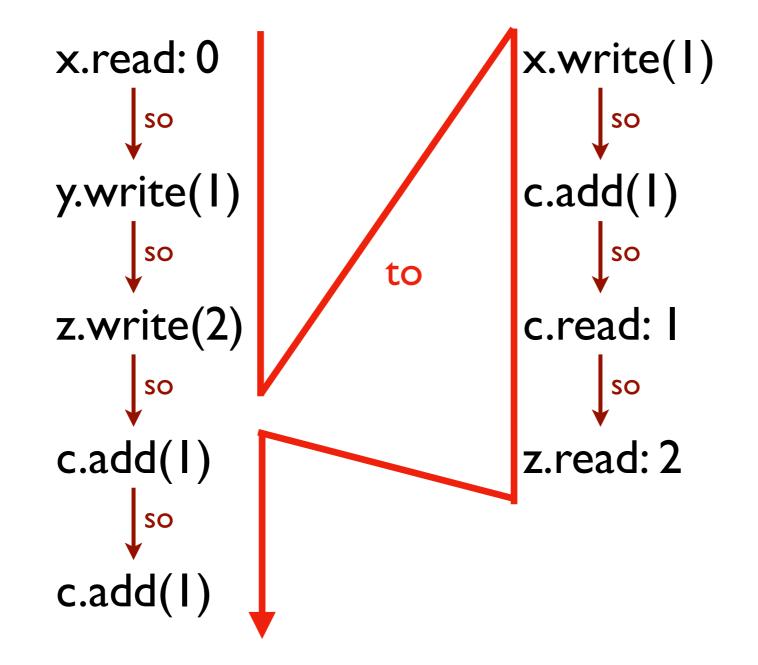
SC example

$$SC = \{(E, so) \mid \exists to. (E, so, to) \vDash \mathscr{A}_{SC}\}$$

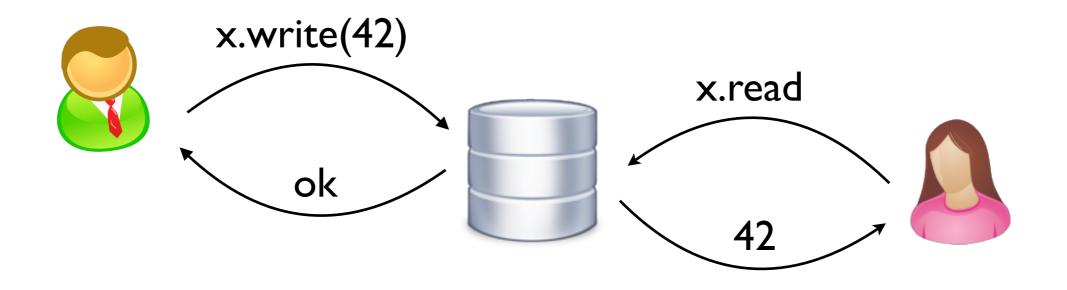


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Operational vs axiomatic

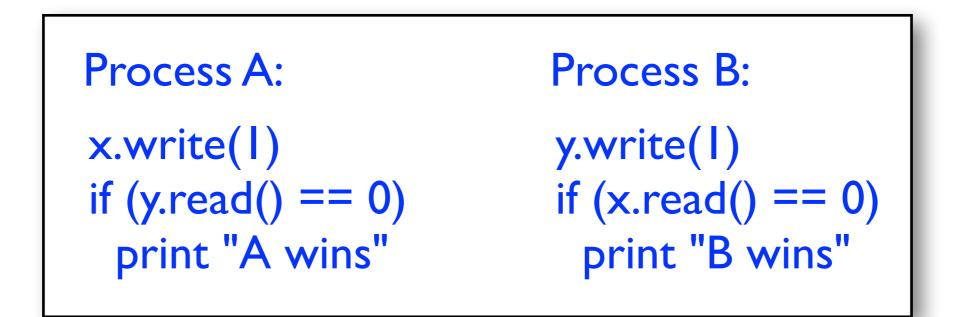


- Got rid of messages between clients and the server, but didn't go far from the operational spec
- There's more difference for weaker models: complex processing can be concisely specified by axioms

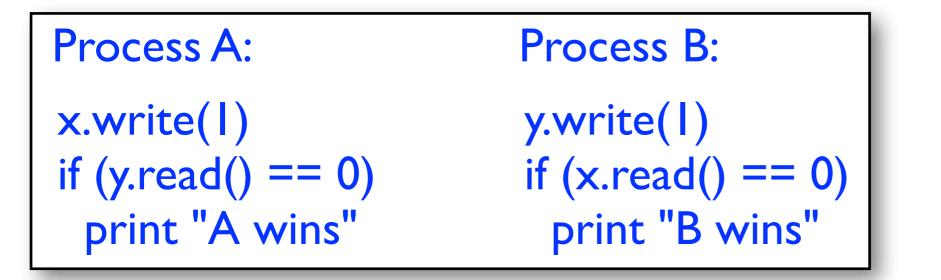
Process A: x.write(1) if (y.read() == 0) print "A wins"

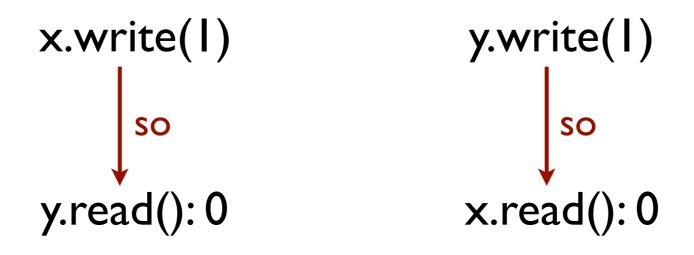
Process B: y.write(1) if (x.read() == 0) print "B wins"

Dekker example

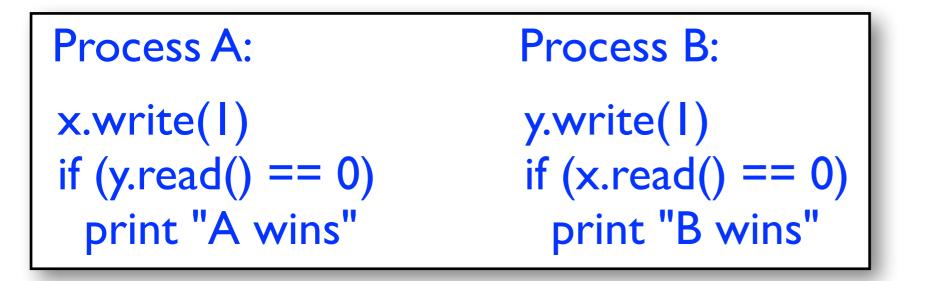


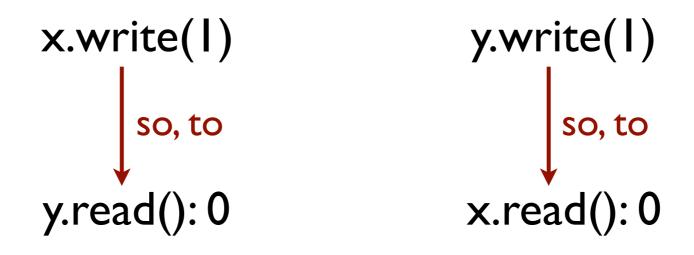
Claim: under sequential consistency, there can be at most one winner

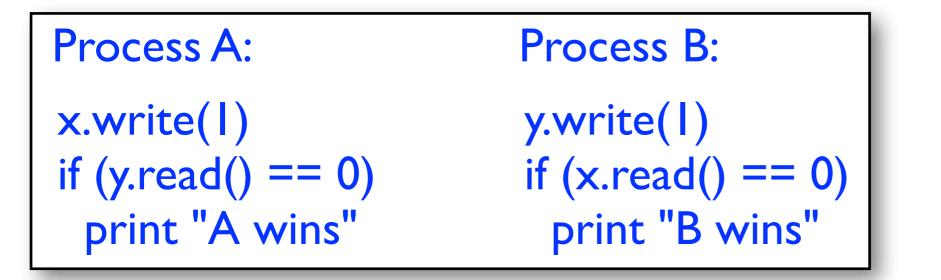


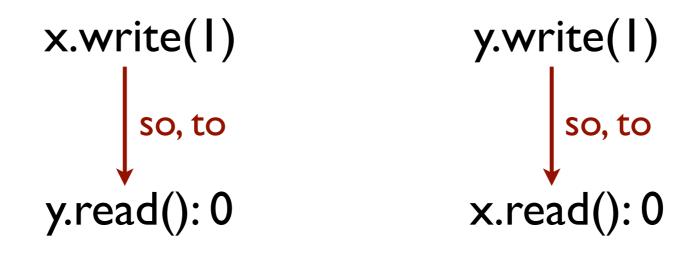


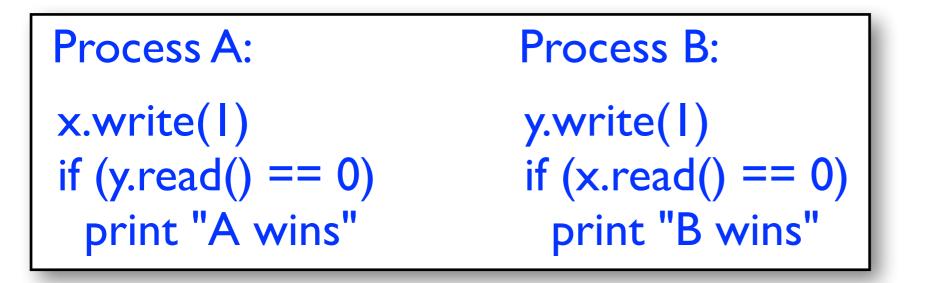
Need to construct a total order to

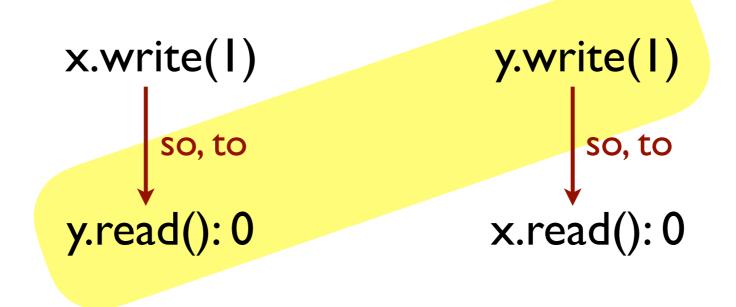


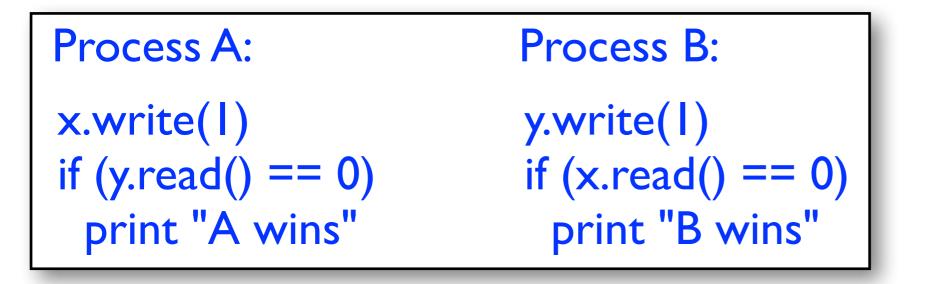


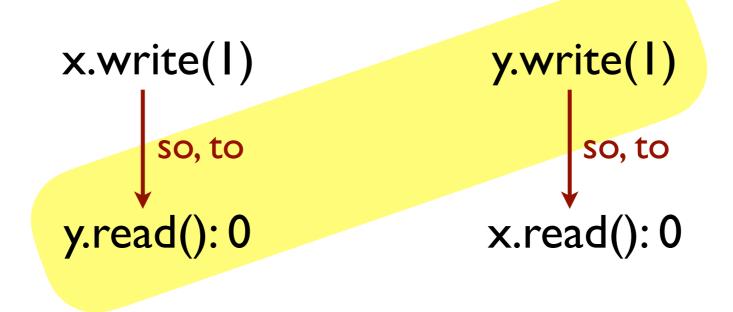


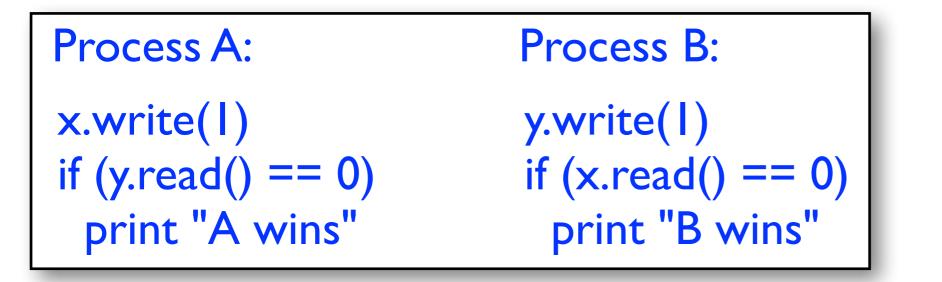


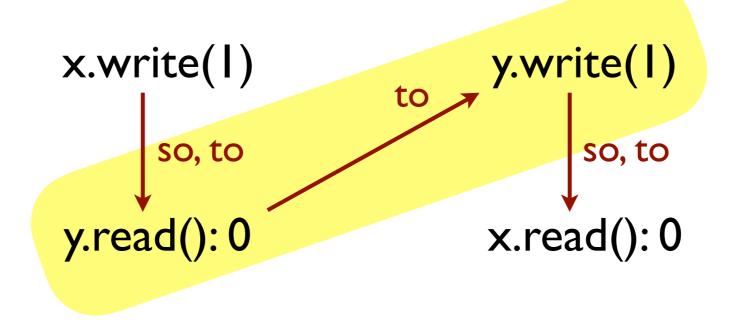


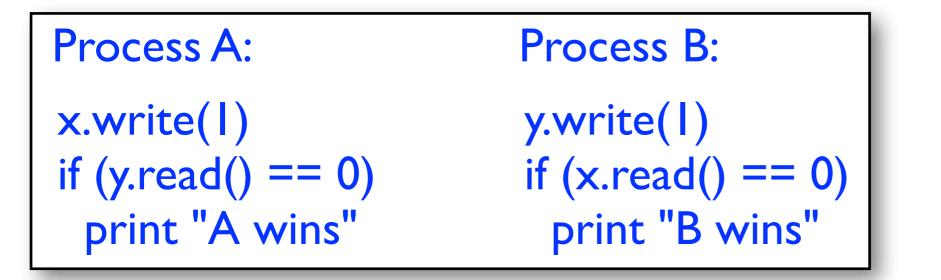


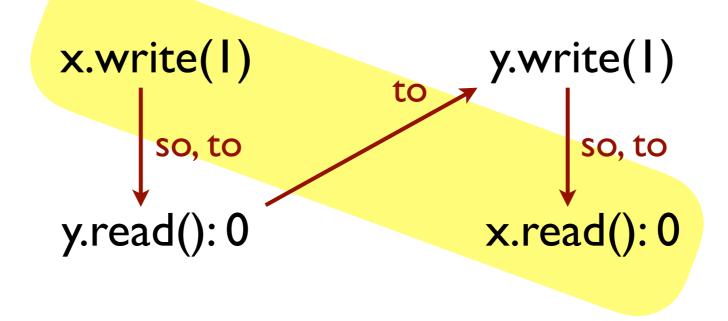


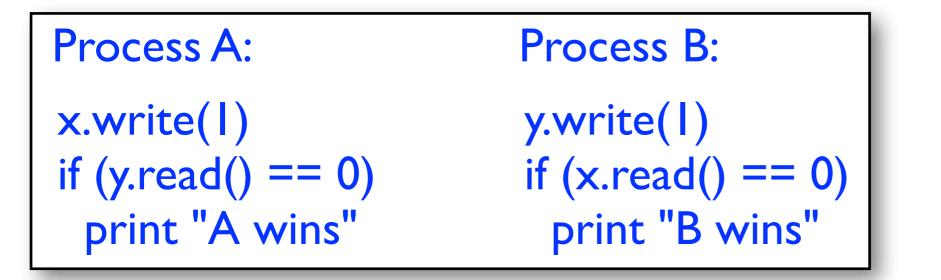


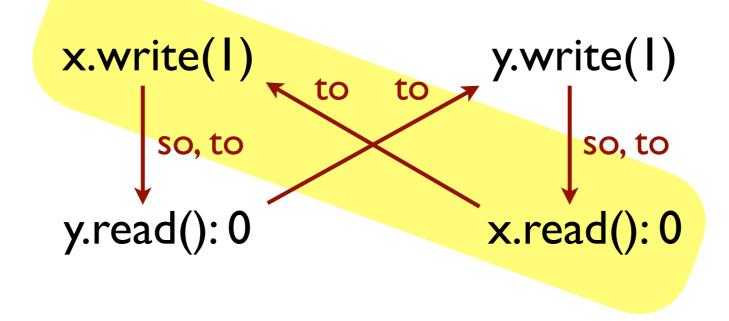


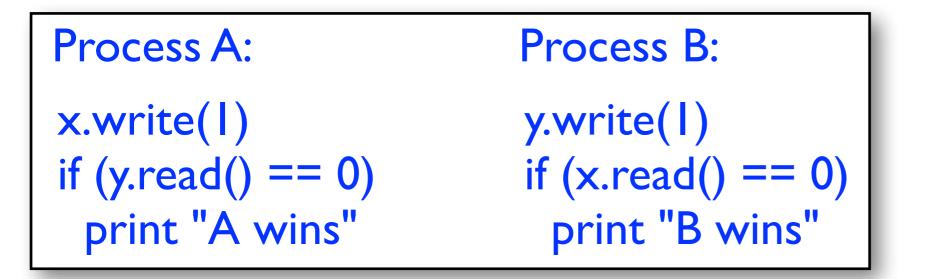


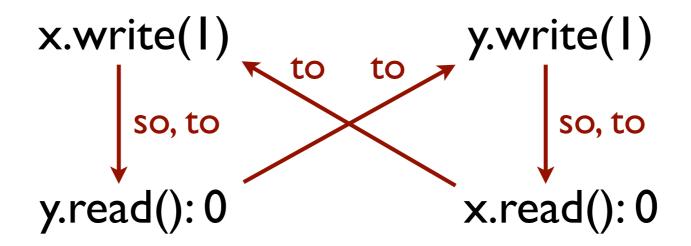












But to must be acyclic, so no such total order exists - QED.

CAP theorem

No system with at least 2 processes can implement a read-write register with strong consistency, availability, and partition tolerance

- strong consistency = sequential consistency
- availability = all operations eventually complete
- partition tolerance = system continues to function under permanent network partitions
 (processes in different partitions can no longer communicate in any way)

CAP proof

No system with at least 2 processes can implement a read-write register with strong consistency, availability, and partition tolerance

- By contradiction: assume the desired system exists
- Run some experiments with the Dekker program
- Network is partitioned between the two processes

Process A:	Process B:
x.write(1)	y.write(1)
if (y.read() == 0)	if $(x.read() == 0)$
print "A wins"	print "B wins"

Process A	Process B
x.write(I) if (y.read() == 0) print "A wins"	

• Process A runs its code, process B is idle

	Process A	Process B
execution X _A of process A	x.write(1) if (y.read() == 0) print "A wins"	

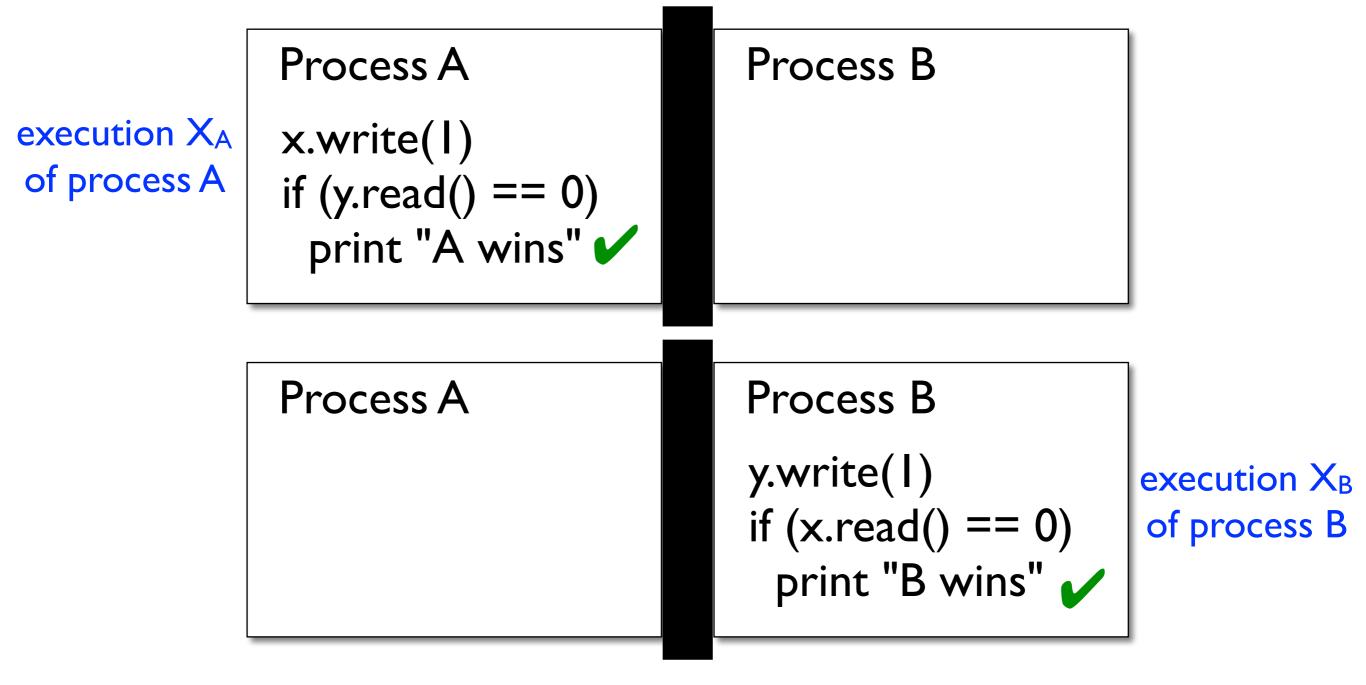
- Process A runs its code, process B is idle
- Availability \Longrightarrow A must terminate and produce an execution X_A

execution X_A
of process A
$$x.write(I)$$

if (y.read() == 0)
print "A wins"

of

- Process A runs its code, process B is idle
- Availability \implies A must terminate and produce an execution X_A
- Sequential consistency $\implies X_A$ must print "A wins"



- Process B runs its code, process A is idle
- Availability \implies B must terminate and produce an execution X_B
- Sequential consistency \implies X_B must print "B wins"

- Network is partitioned in both experiments: processes didn't receive any messages
- X_A; X_B is an execution of A || B, i.e., Dekker
- $X_A; X_B \text{ not } SC \Longrightarrow \text{ contradiction, } QED$

 Processes have to talk to each other (synchronise) to guarantee strong consistency Eventual consistency and replicated data types, operationally

System model

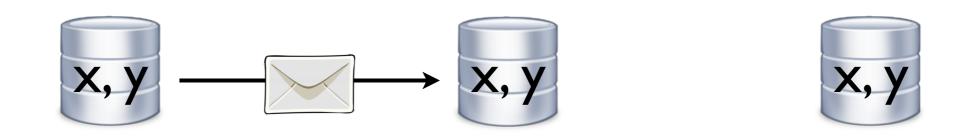






- Database system consisting of multiple replicas (= data centre, machine, mobile device)
- Each replica stores a copy of all objects

System model



- Replicas can communicate via channels
- Asynchronous: no bound on how quickly a message will be delivered
 (in particular, because of network partitions)
- Reliable: every message is eventually delivered (so every partition eventually heals)
- For now: replicas are reliable too

High availability







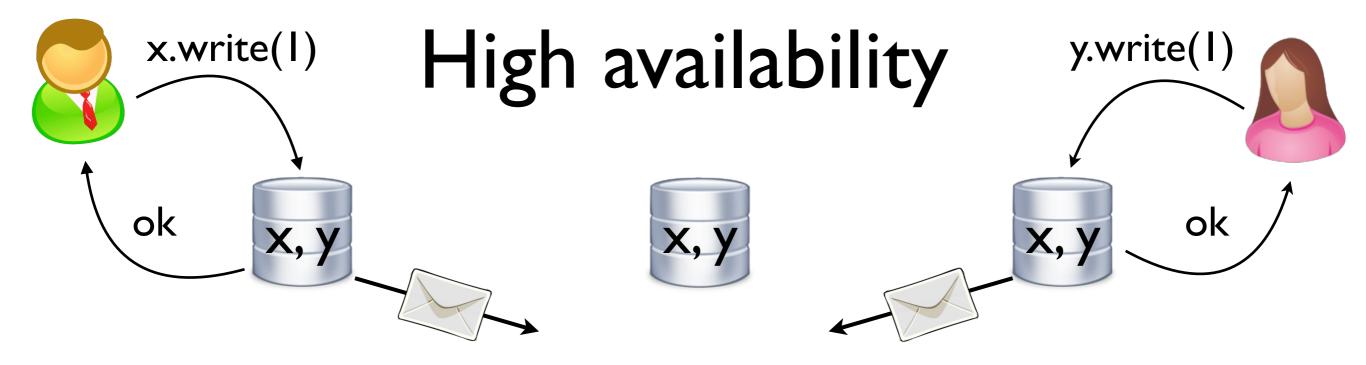
• Clients connect to a replica of their choice



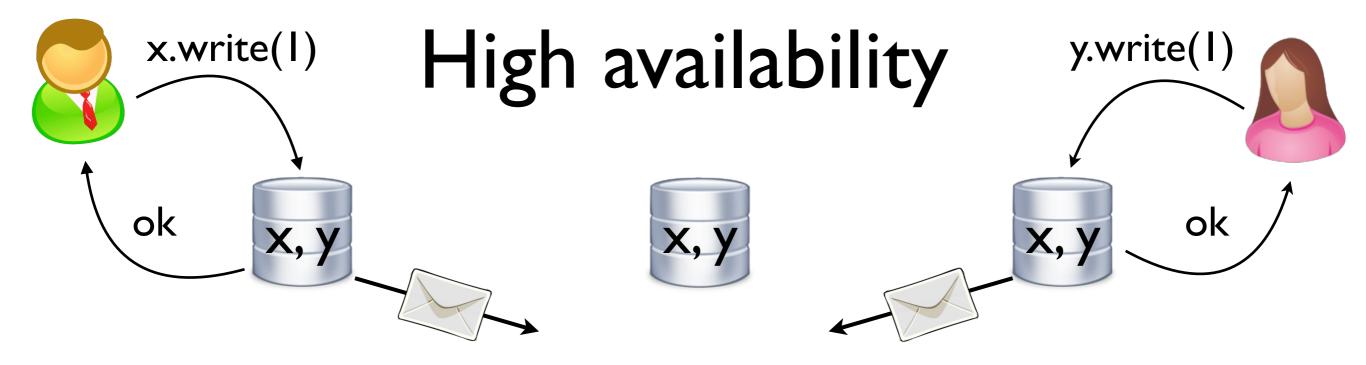
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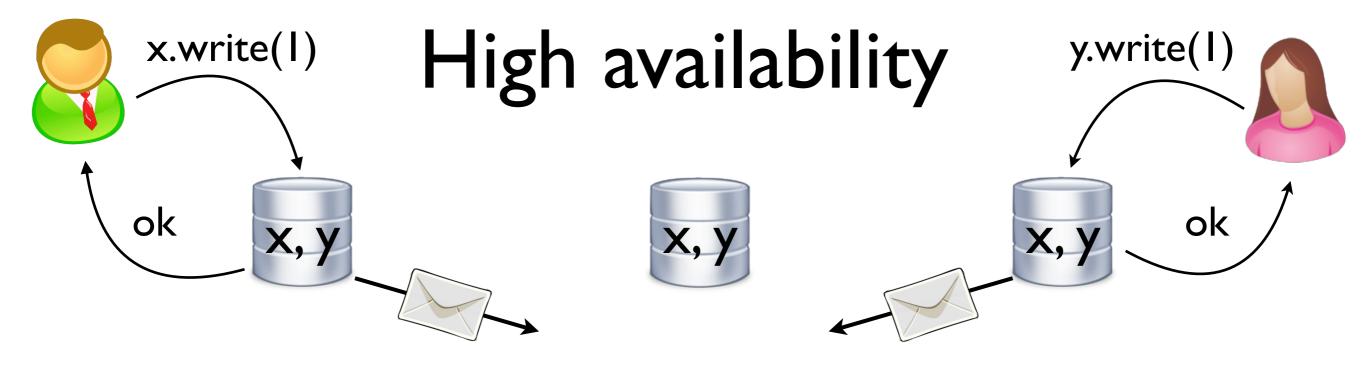
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- Replica has to respond to operations immediately, without communicating with others



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- Propagate effects to other replicas later



- Clients connect to a replica of their choice
- Replica has to respond to operations immediately, without communicating with others
- Propagate effects to other replicas later
- Always available, low latency, but may not be strongly consistent



- Quiescent consistency: if no new updates are made to the database, then replicas will eventually converge to the same state
- Later more precise and stronger formulations of eventual consistency

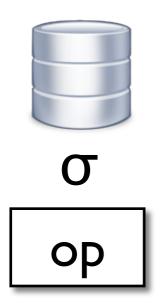
- Need a new kind of replicated data type: object state now lives at multiple replicas
- Aka CRDTs: commutative, convergent, conflict-free
 Just one type: operation-based replicated data types
- Object → Type → Operation signature
 For now fix a single object and type

Sequential semantics recap

- Set of states State
- Initial state $\sigma_0 \in State$
- $[op]_{val} \in State \rightarrow Value$
- $[op]_{state} \in State \rightarrow State$

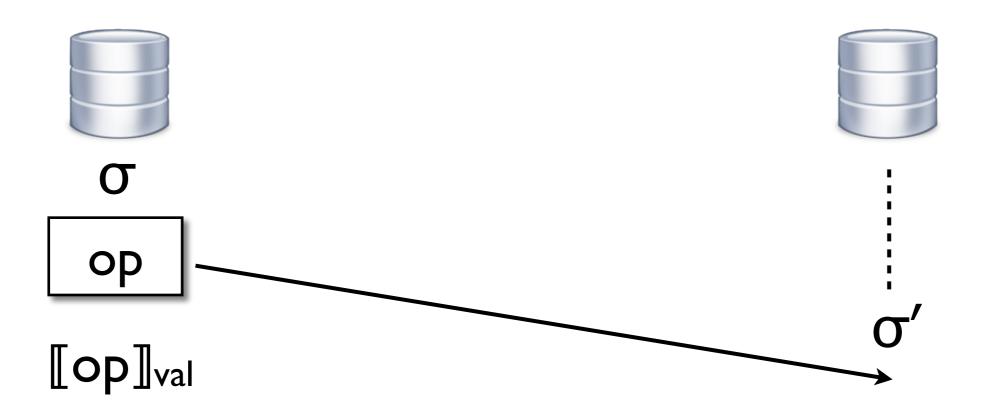


Object state at a replica: $\sigma \in State$

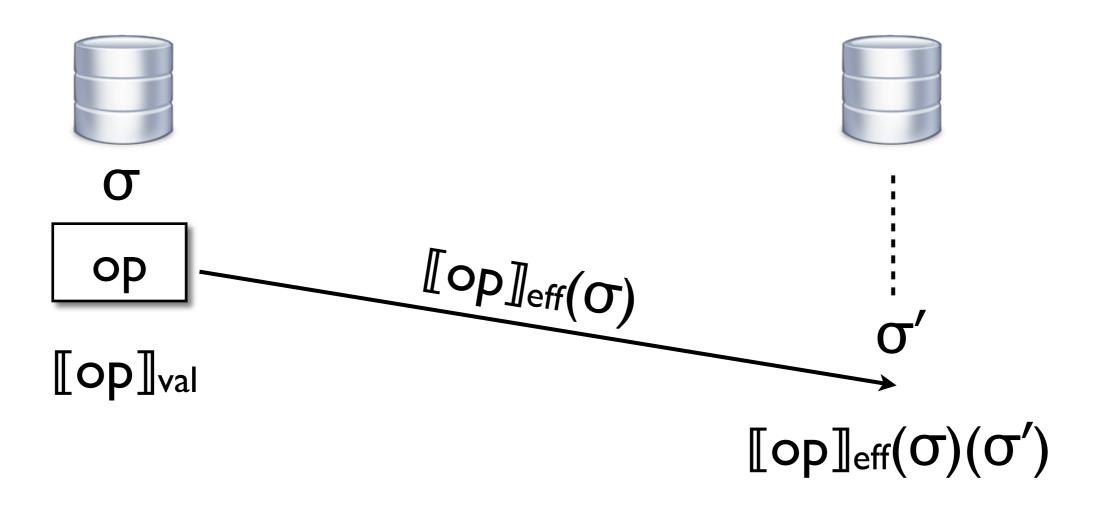




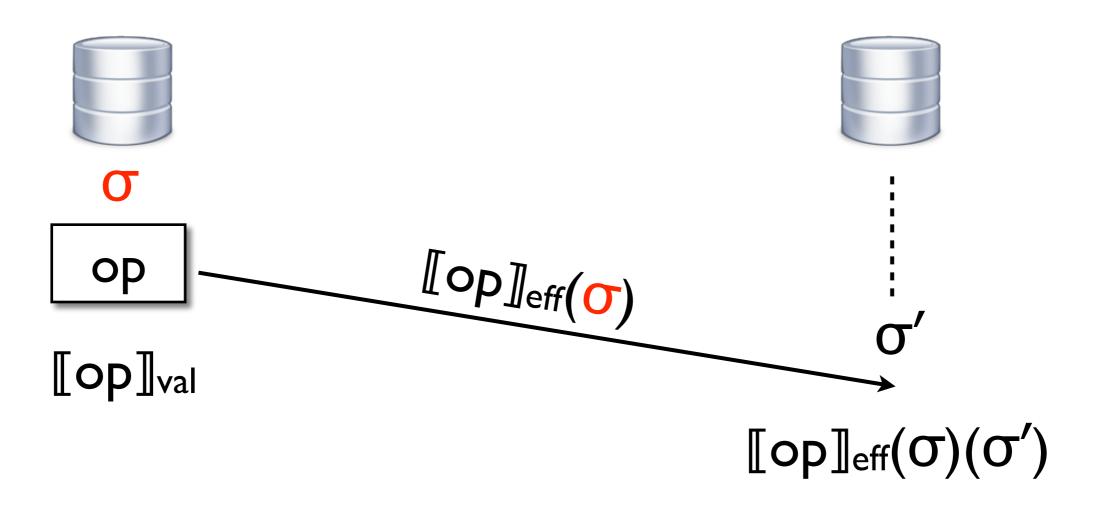
Object state at a replica: $\sigma \in State$ Return value: $[op]_{val} \in State \rightarrow Value$



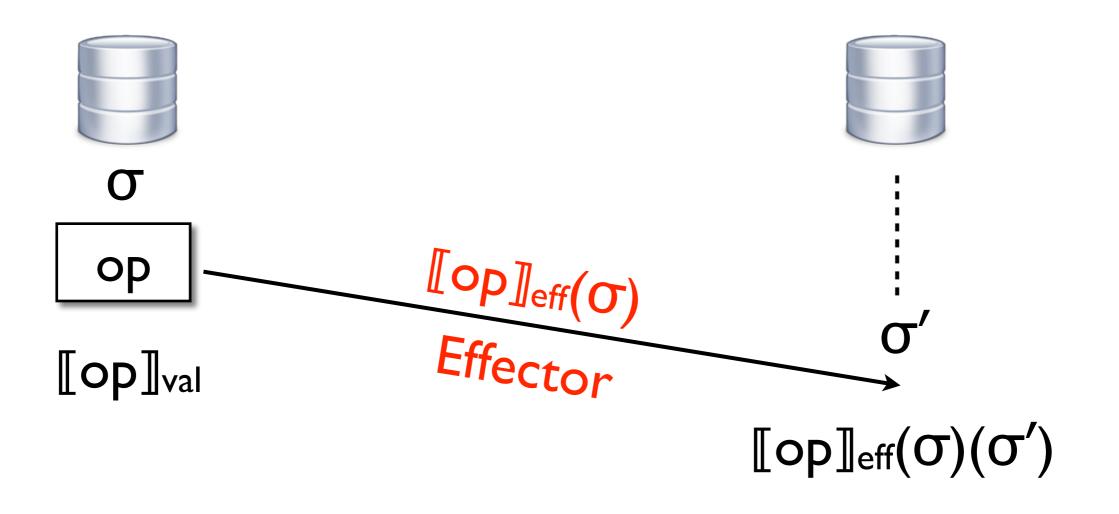
Object state at a replica: $\sigma \in State$ Return value: $[\sigma p]_{val} \in State \rightarrow Value$ The operation affects a different state σ' !



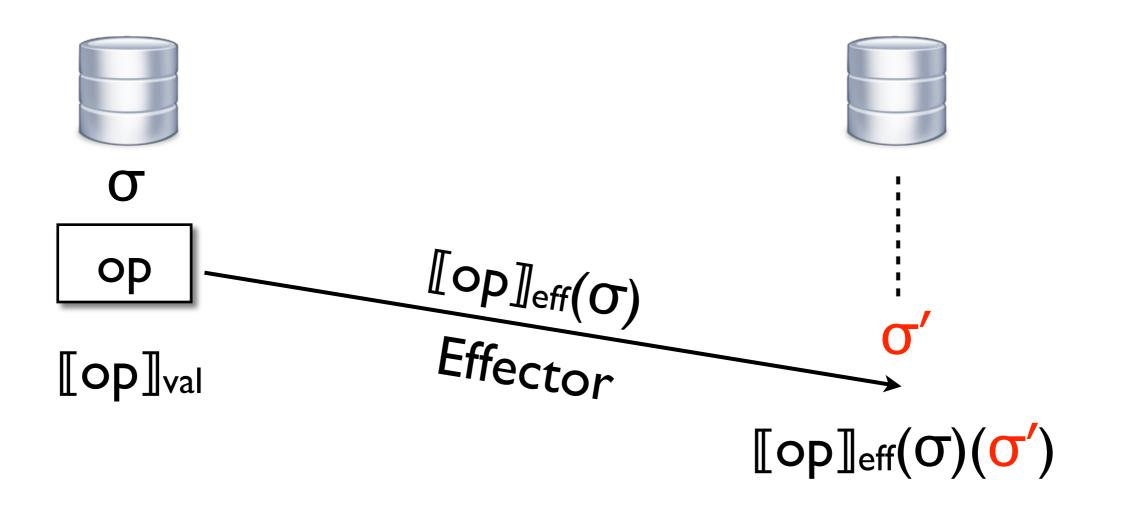
Object state at a replica: $\sigma \in \text{State}$ Return value: $[\sigma p]_{val} \in \text{State} \rightarrow \text{Value}$ Effector: $[\sigma p]_{eff} \in \text{State} \rightarrow (\text{State} \rightarrow \text{State})$



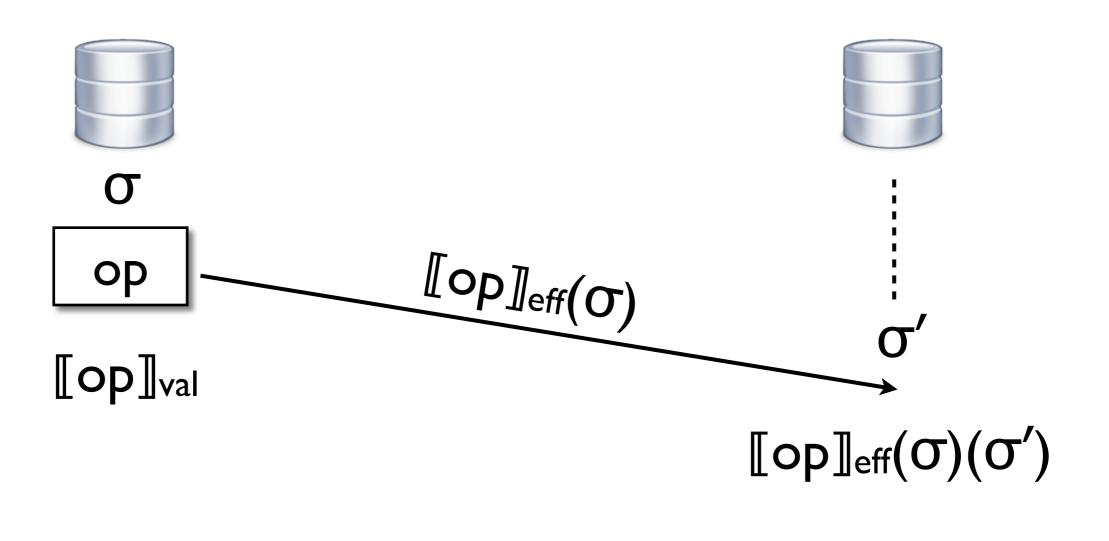
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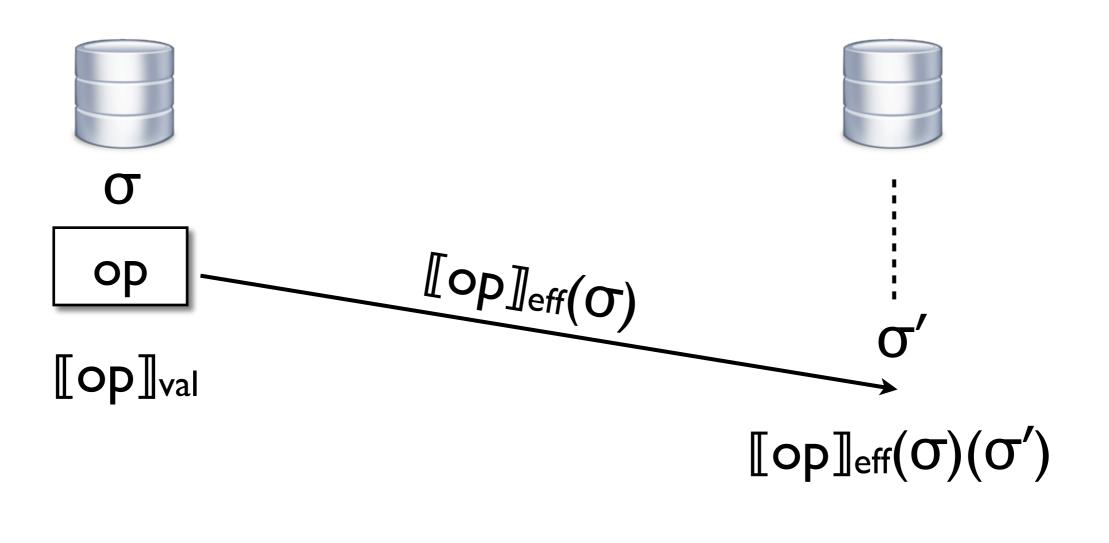
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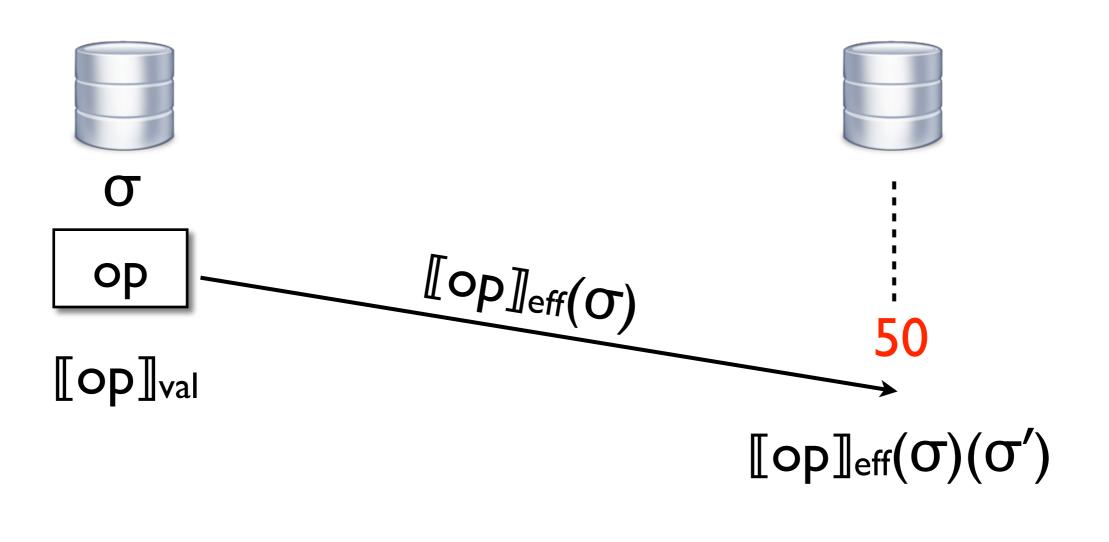
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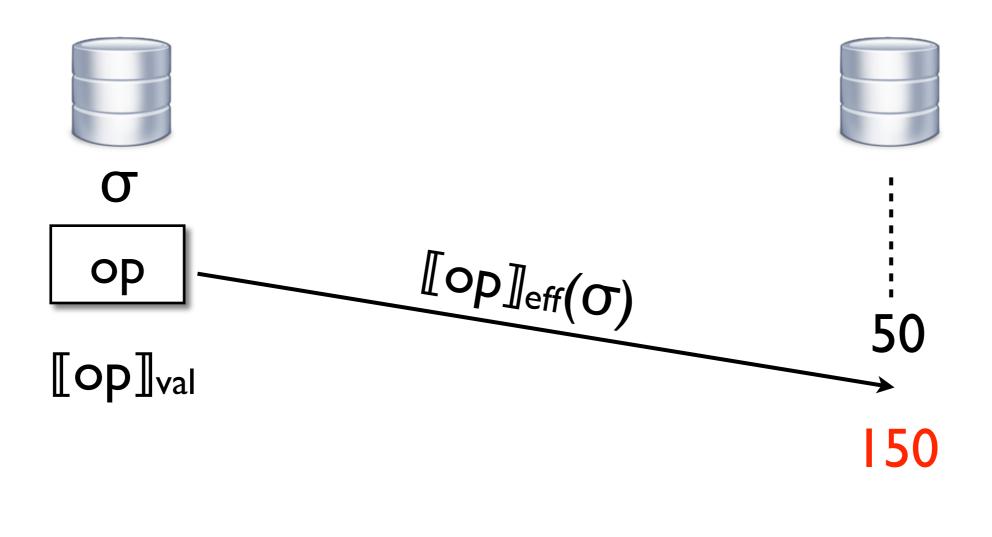
State = \mathbb{N} [read()]_{val}(σ) = σ [read()]_{eff}(σ) = $\lambda \sigma$. σ



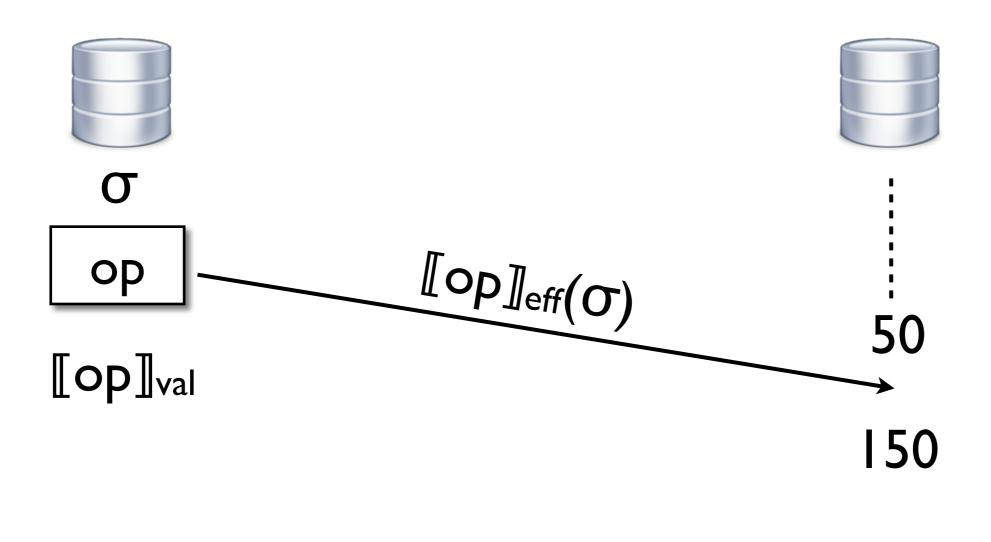
 $[add(100)]_{eff}(\sigma) = \lambda\sigma'.(\sigma' + 100)$



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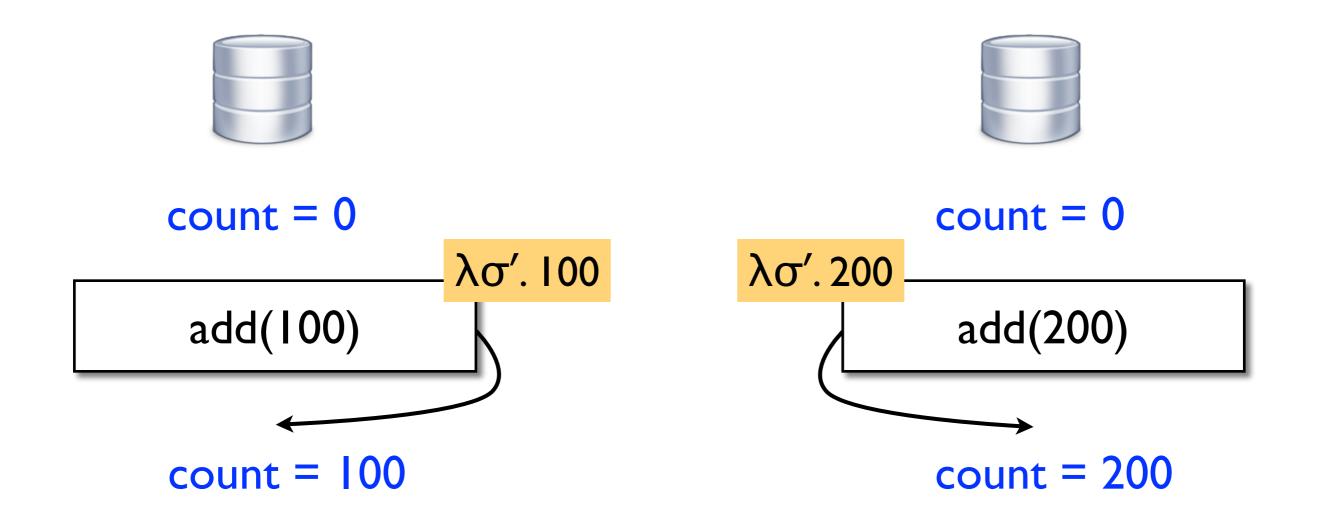
count = 0

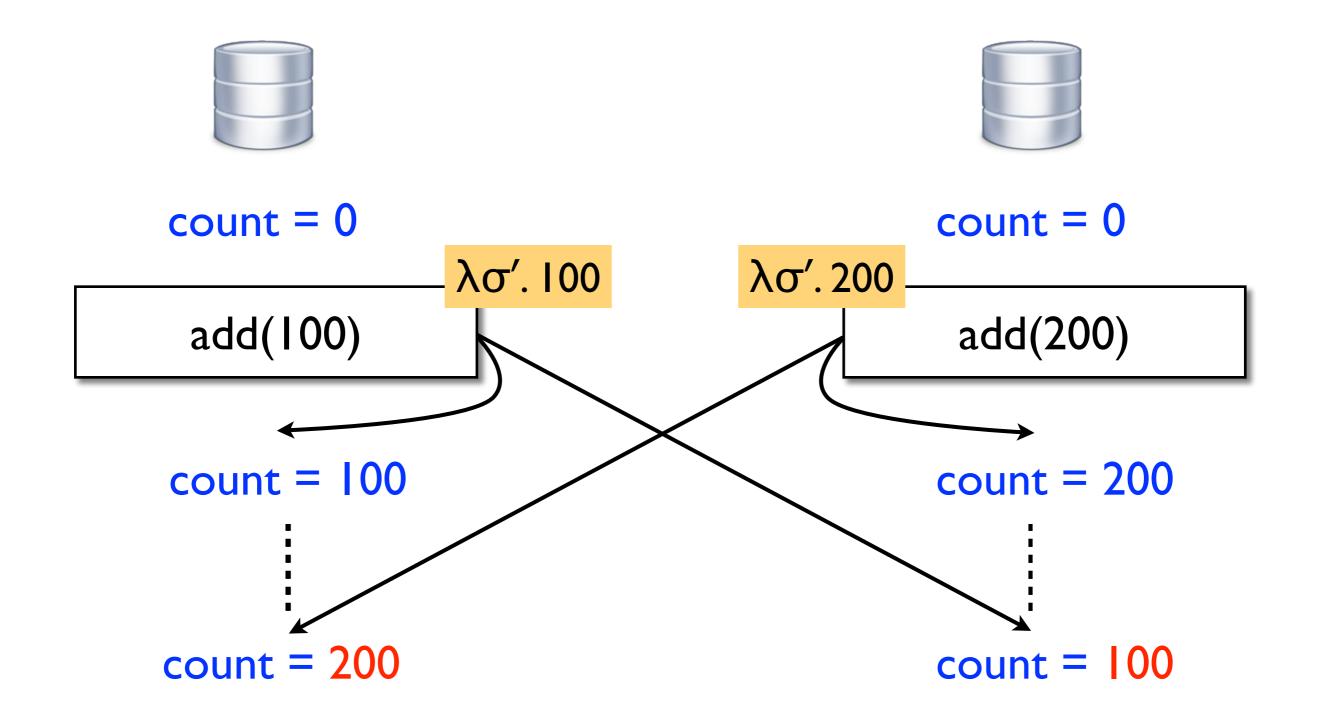
add(100)



count = 0

add(200)





Quiescent consistency violated: all updates have been delivered, yet replicas will never converge

Ensuring quiescent consistency

• Effectors have to commute:

 $\forall op_{1}, op_{2}, \sigma_{1}, \sigma_{2}. \ [op_{1}]_{eff}(\sigma_{1}); \ [op_{2}]_{eff}(\sigma_{2}) = \\ [op_{2}]_{eff}(\sigma_{2}); \ [op_{1}]_{eff}(\sigma_{1}) \end{cases}$

 Convergence: replicas that received the same sets of updates end up in the same state
 (even when messages are received in different orders)

Ensuring quiescent consistency

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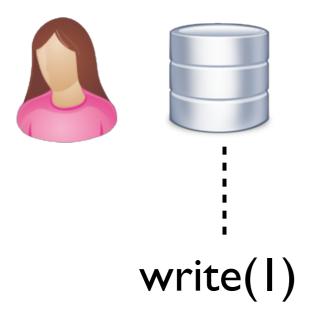
$$\forall op_1, op_2, \sigma_1, \sigma_2. [op_1]_{eff}(\sigma_1); [op_2]_{eff}(\sigma_2) = [op_2]_{eff}(\sigma_2); [op_1]_{eff}(\sigma_1)$$

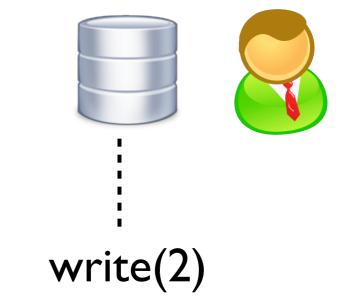
- Convergence: replicas that received the same sets of updates end up in the same state
 (even when messages are received in different orders)
- Quiescent consistency: if no new updates are made to the database, then replicas will eventually converge to the same state

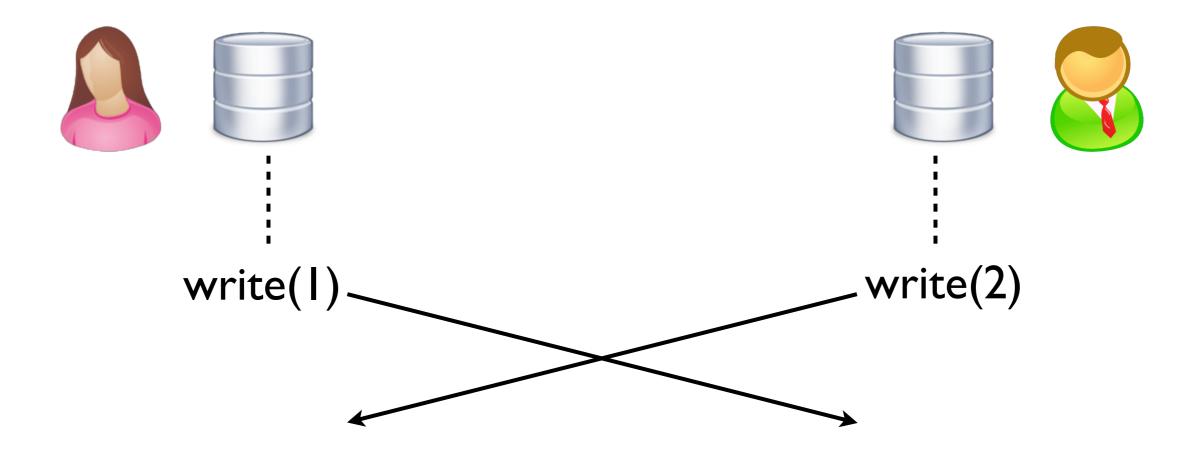
(because update get eventually delivered)

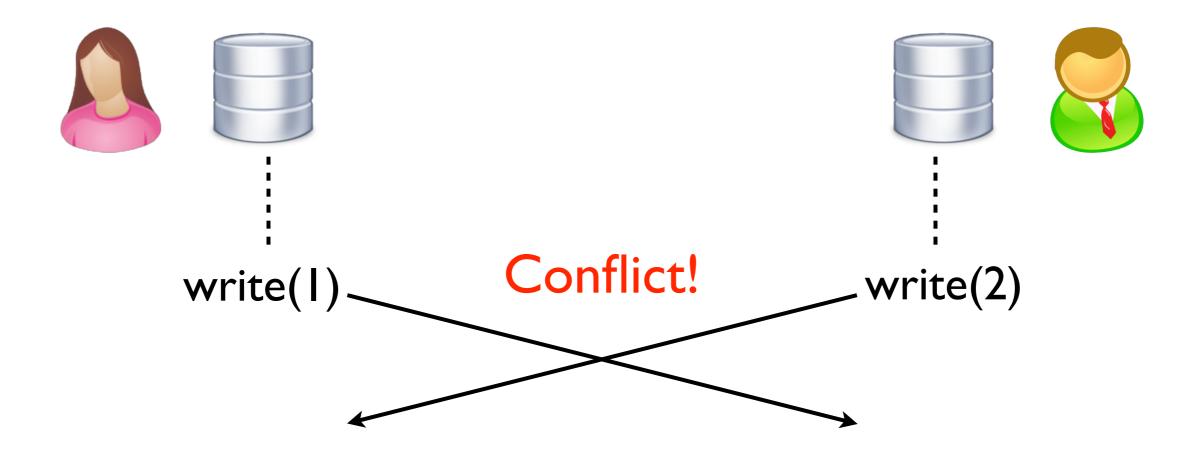
• Counter

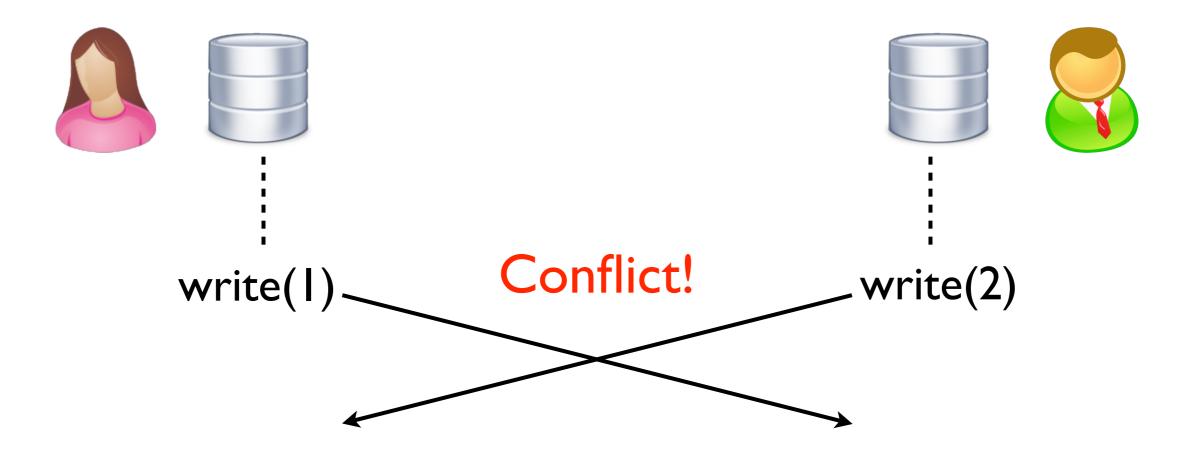
- Last-writer-wins register
- Multi-valued register
- Add-wins set
- Remove-wins set
- List
- ...



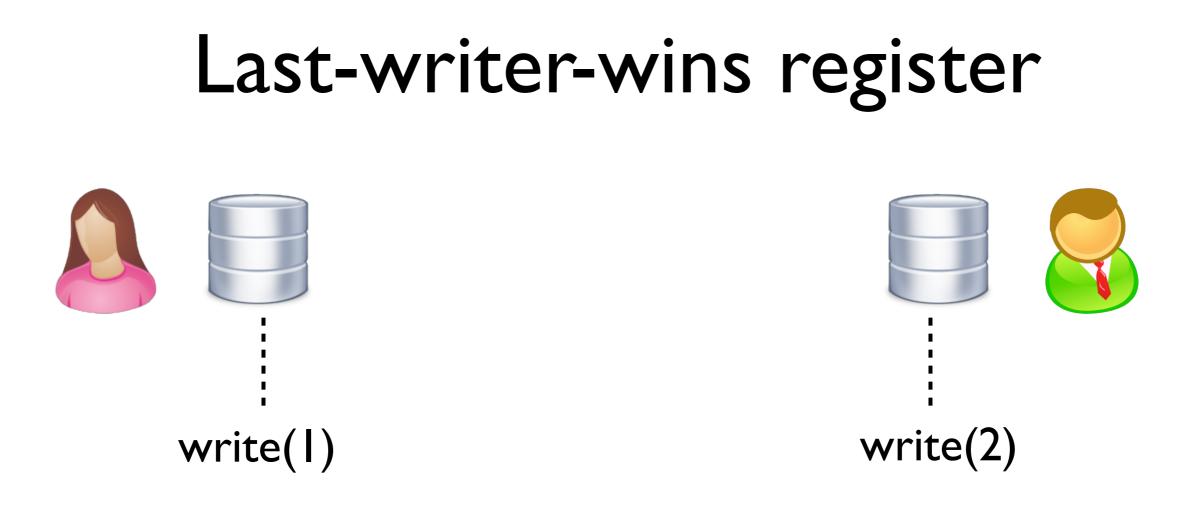








- No right or wrong solutions: depends on the application requirements
- E.g., could report the conflict to the user: multi-valued register



- Shared memory: an arbitrary write will win
- Conflict arbitrated using timestamps: last write wins
- Link to shared-memory consistency models

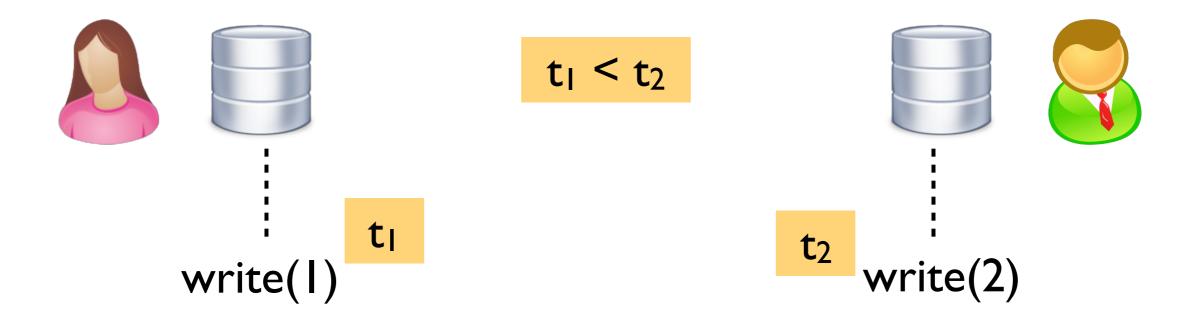


State = Value × Timestamp

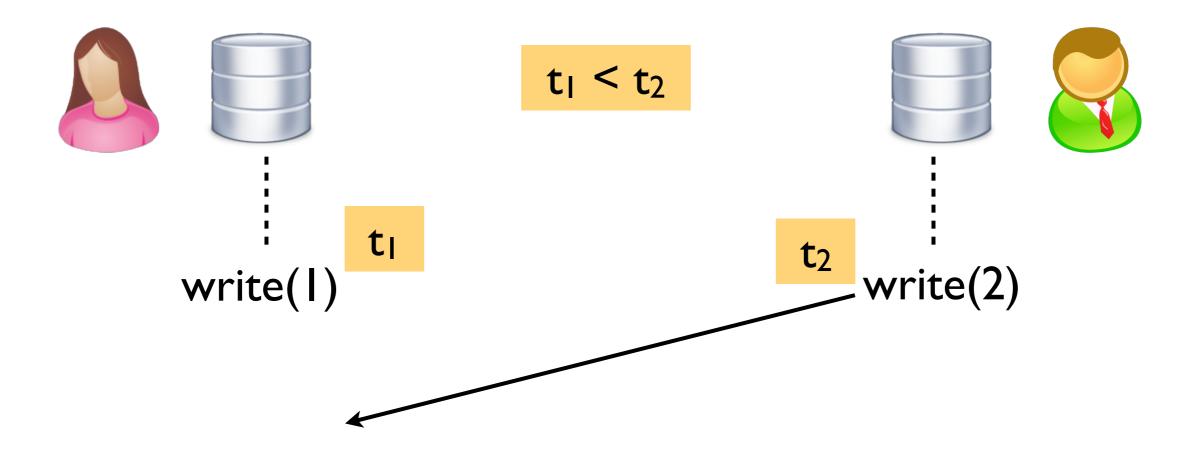
 $[[read()]]_{val}(v, t) = v$



 $[[write(v_{new})]]_{eff}(v, t) =$ $let t_{new} = newUniqueTS() in$ $\lambda(v', t'). if t_{new} > t' then (v_{new}, t_{new}) else (v, t)$

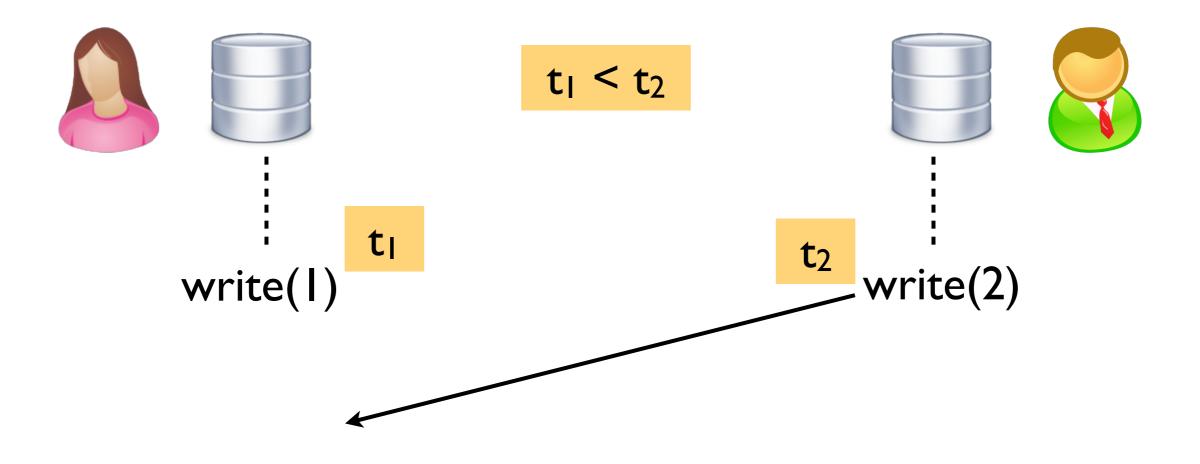


 $[vrite(v_{new})]_{eff}(v, t) =$ $let t_{new} = newUniqueTS() in$ $\lambda(v', t'). if t_{new} > t' then (v_{new}, t_{new}) else (v, t)$



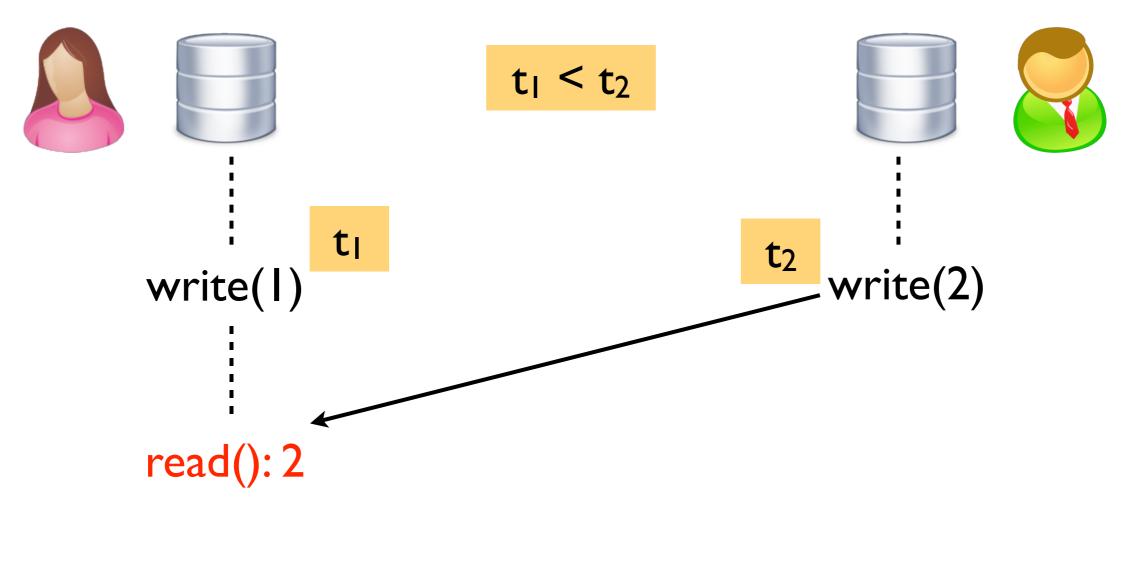
 $[vrite(v_{new})]_{eff}(v, t) =$

let $t_{new} = newUniqueTS()$ in $\lambda(v', t')$. if $t_{new} > t'$ then (v_{new}, t_{new}) else (v, t)



 $[[write(v_{new})]]_{eff}(v, t) =$

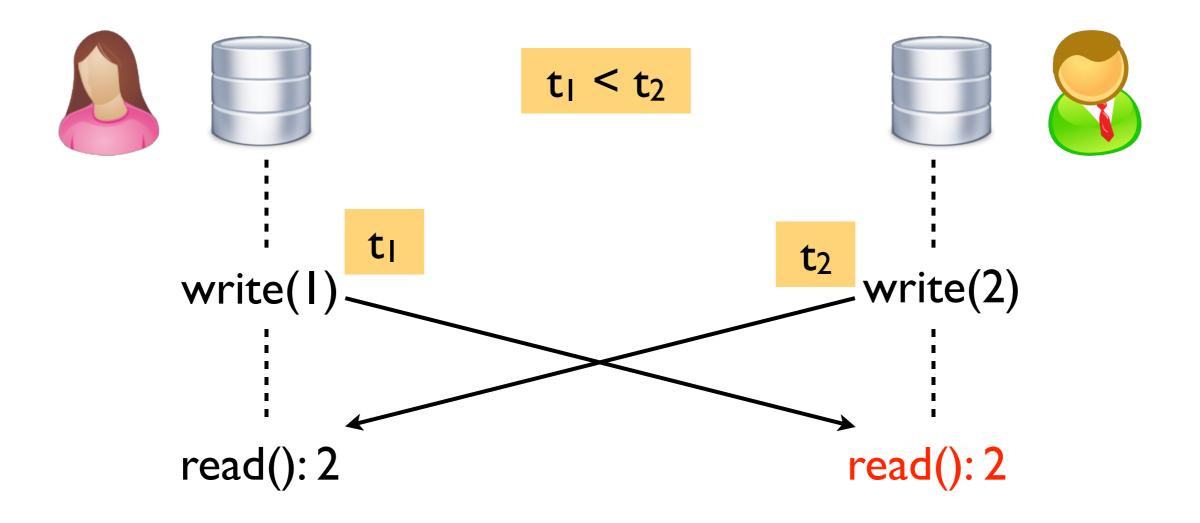
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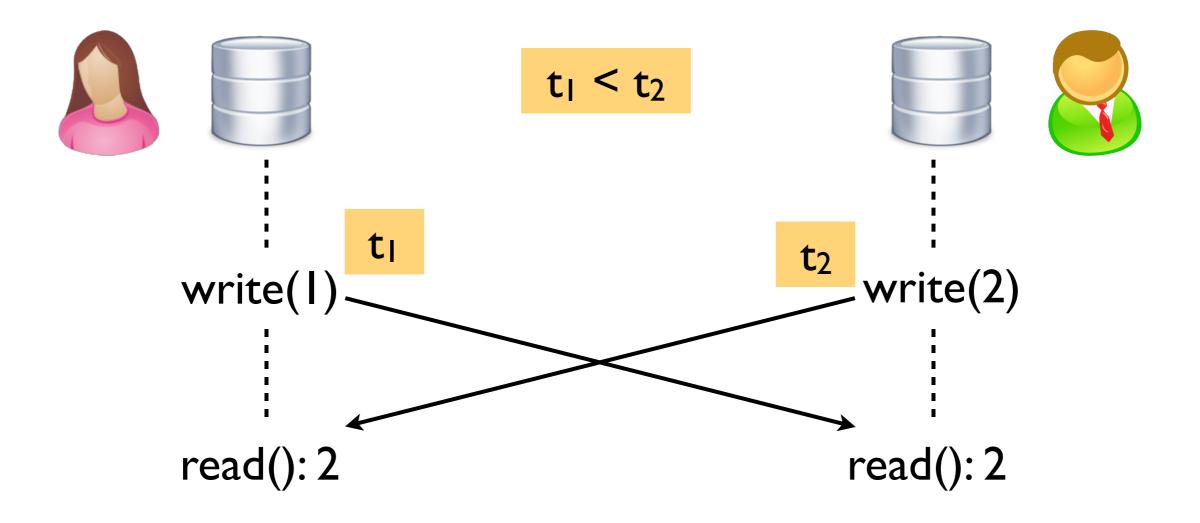
Last-writer-wins register



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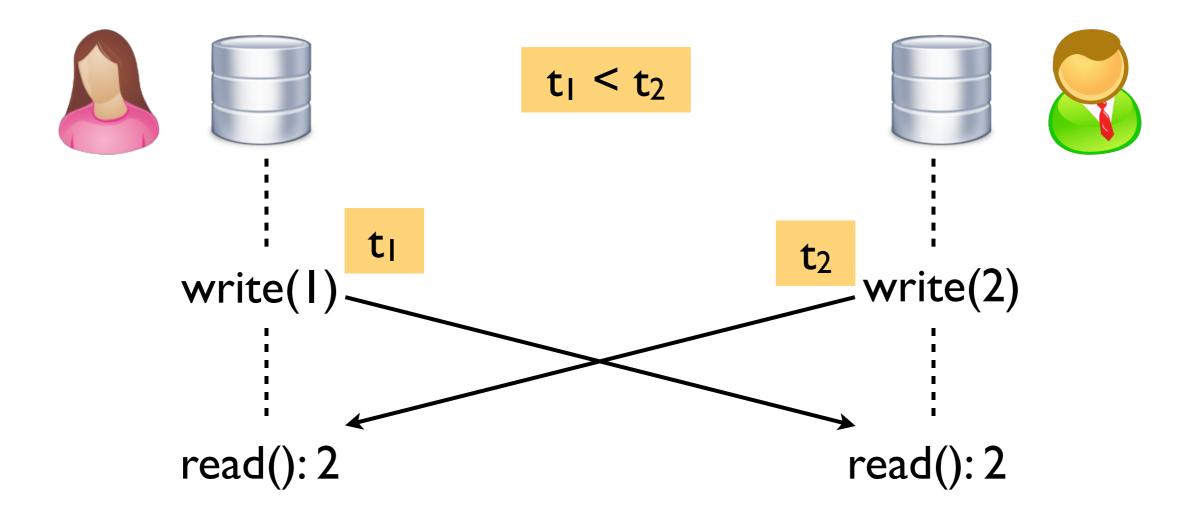
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 $[[write(v_{new})]]_{eff}(v, t) =$

let $t_{new} = newUniqueTS()$ in $\lambda(v', t')$. if $t_{new} > t'$ then (v_{new}, t_{new}) else (v, t)

Last-writer-wins register

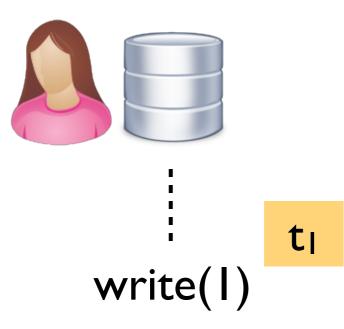


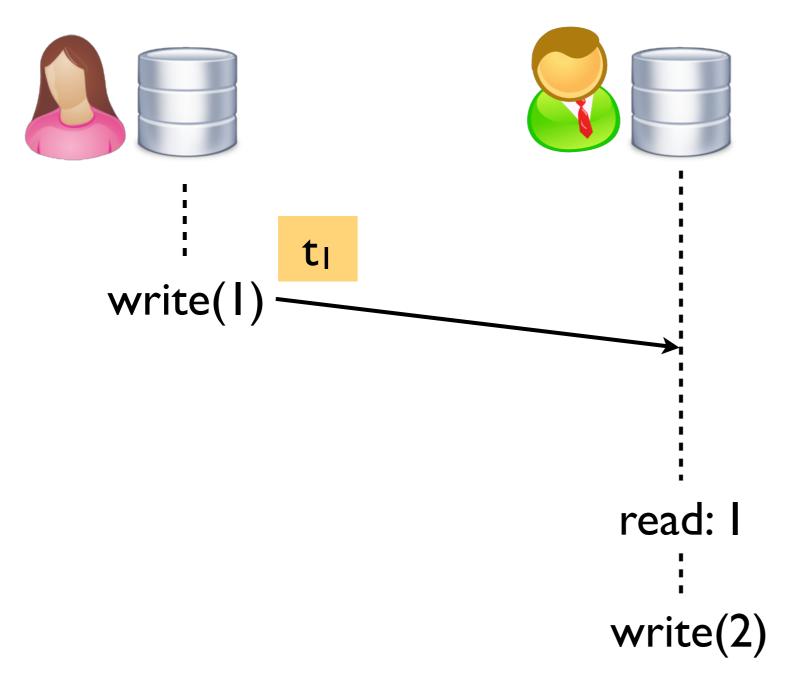
Effectors are commutative: the write with the highest timestamp wins regardless of the order of application

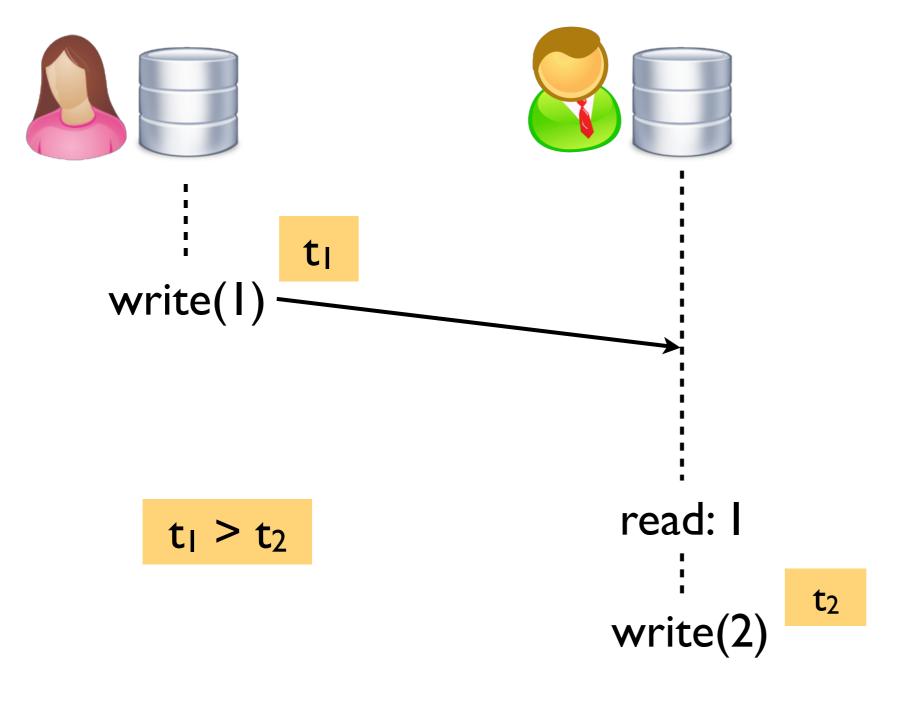
Generating timestamps

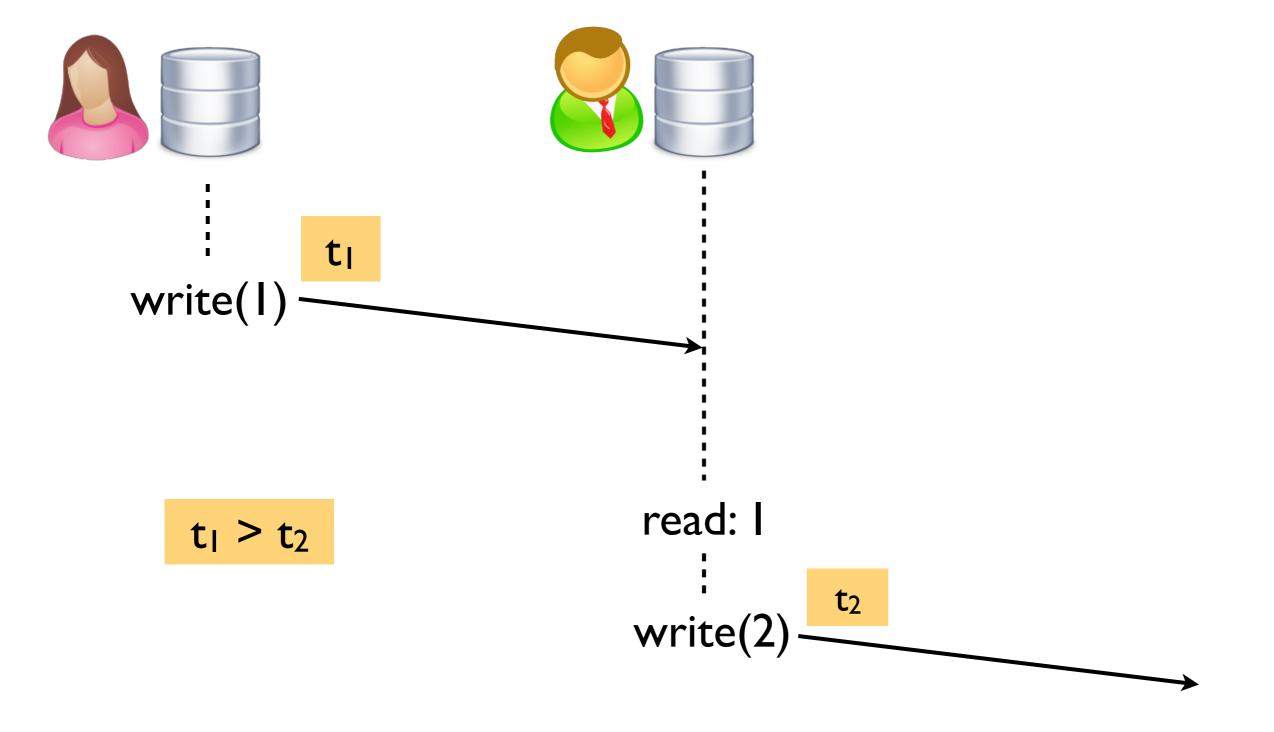
• Can use wall-clock time at the machine

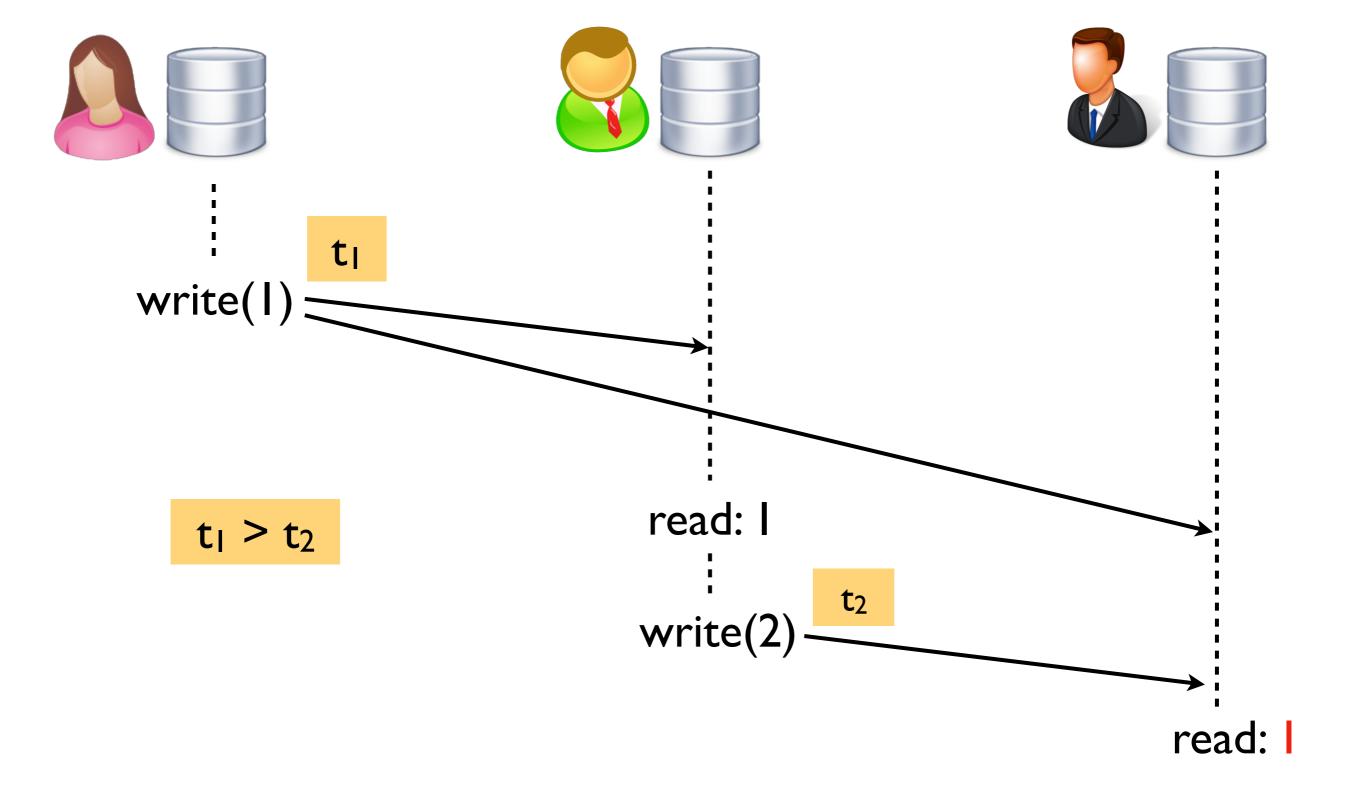
But can lead to strange results when clocks are out of sync

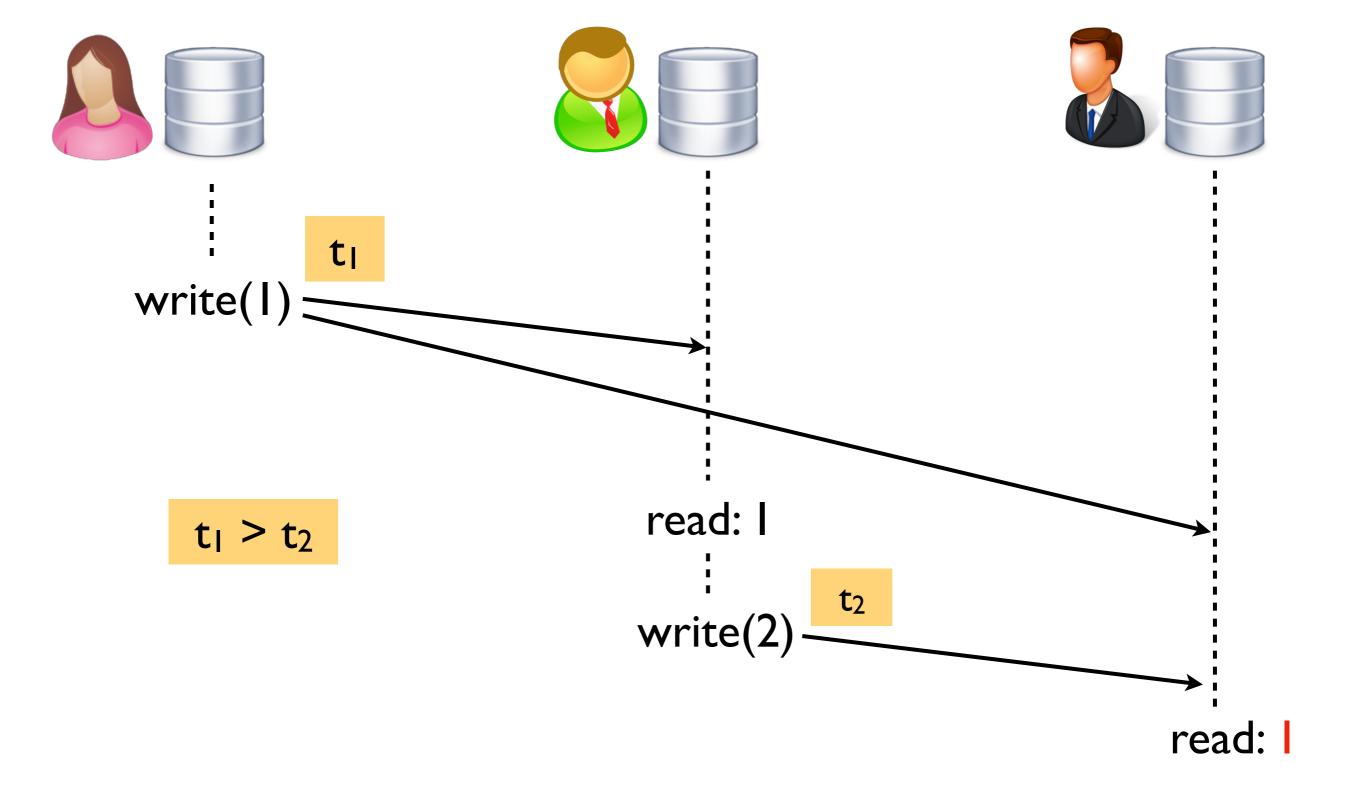




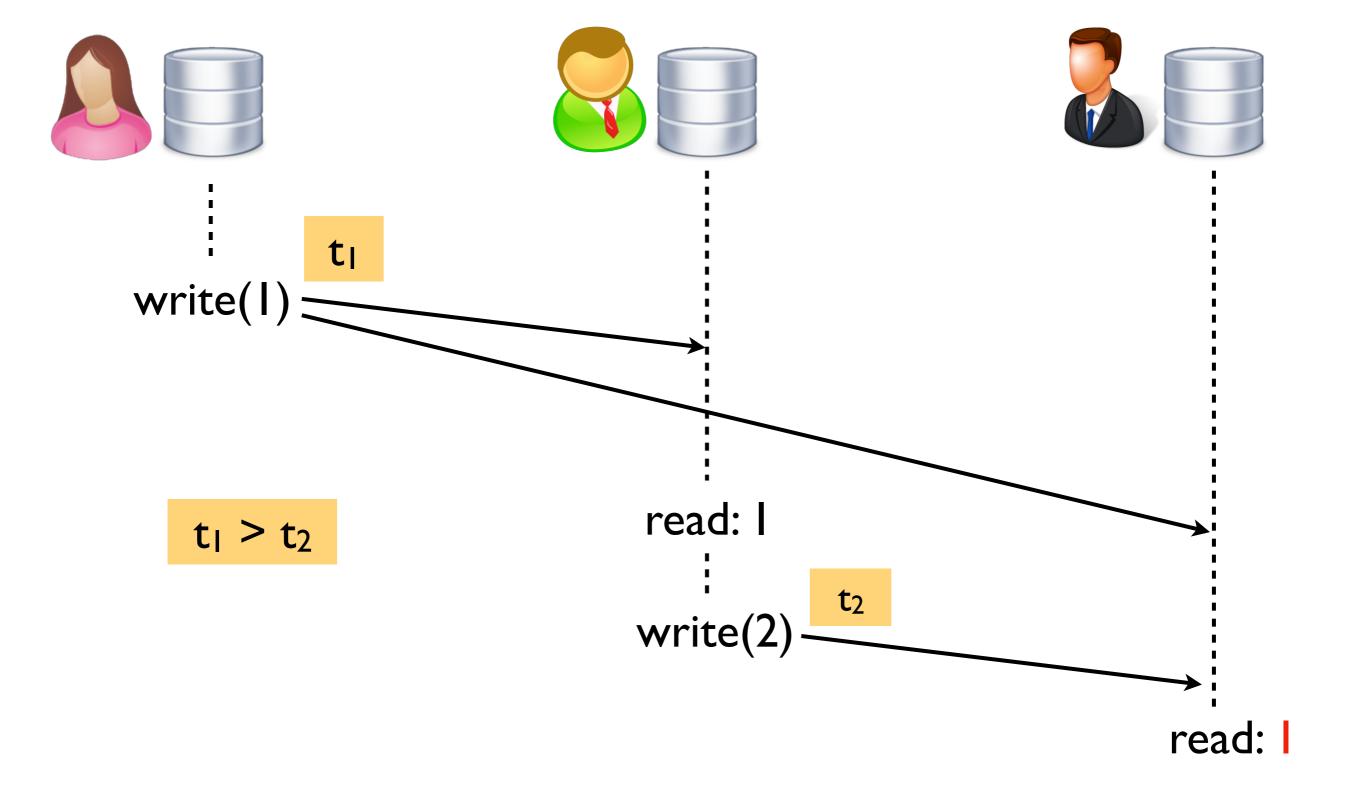






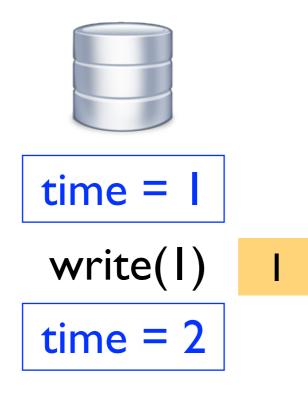


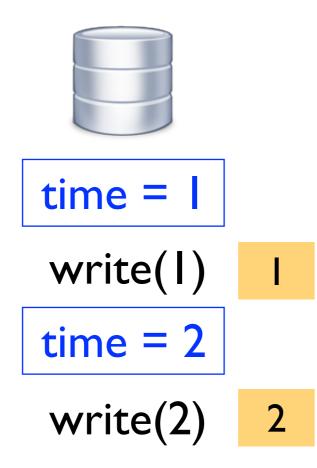
• Undesirable: 2 was meant to supersede I

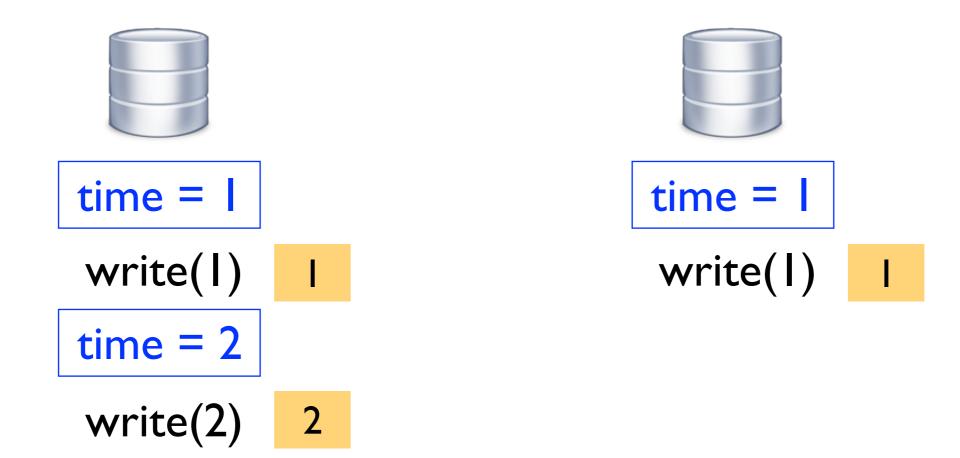


- Undesirable: 2 was meant to supersede I
- Use logical (Lamport) clocks instead

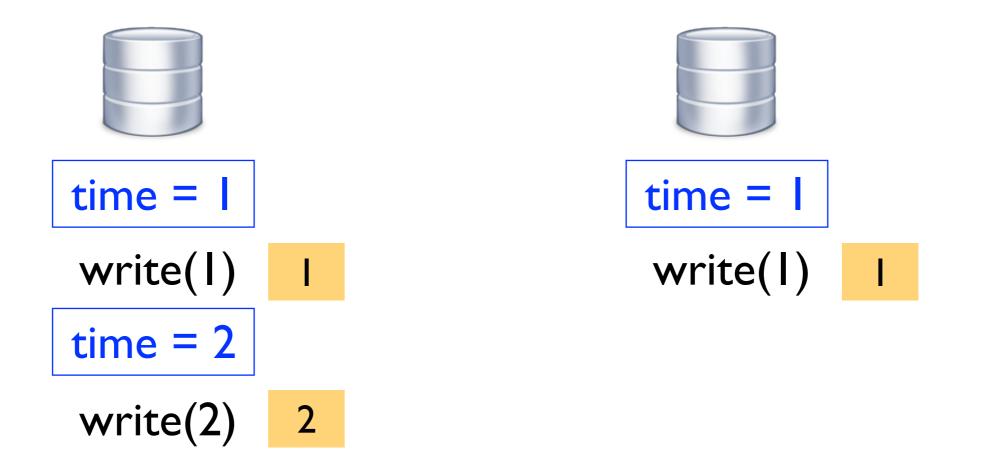






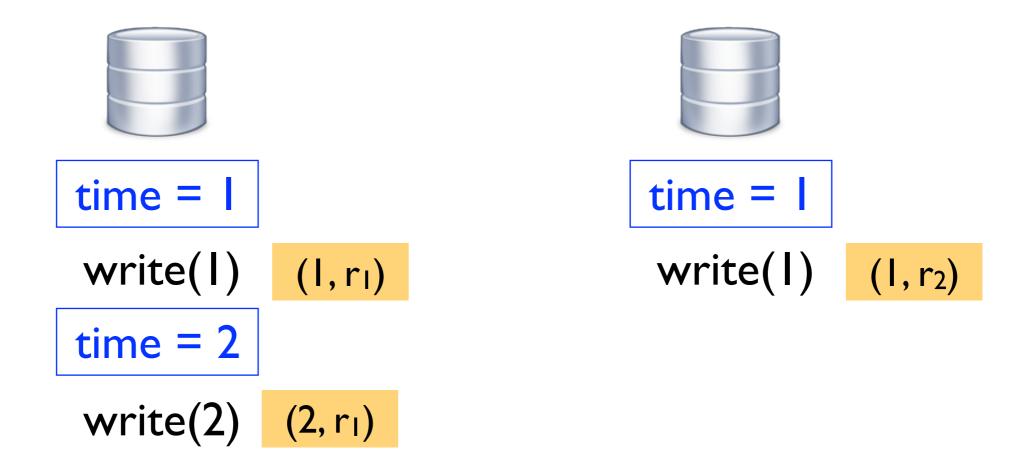


Replica maintains a counter, incremented on each operation:



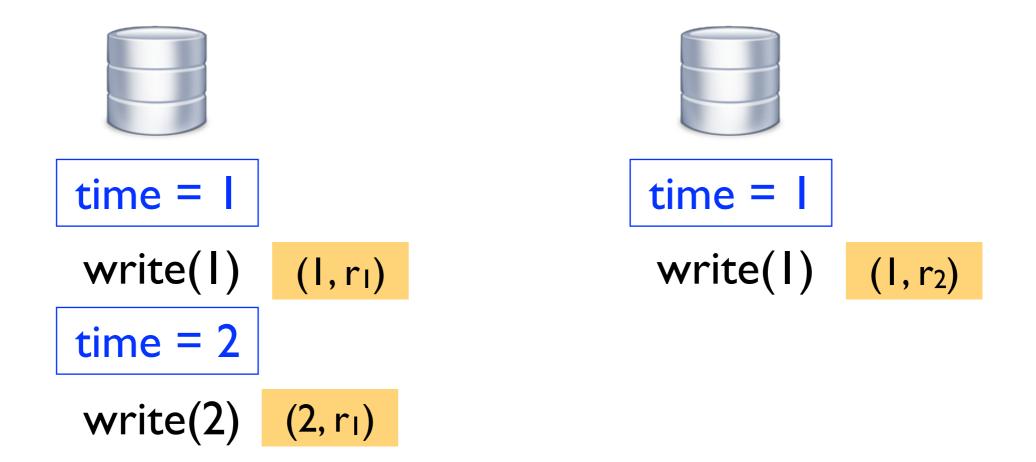
Timestamps need to be unique: ts = (CounterValue, ReplicaID)

Replica maintains a counter, incremented on each operation:



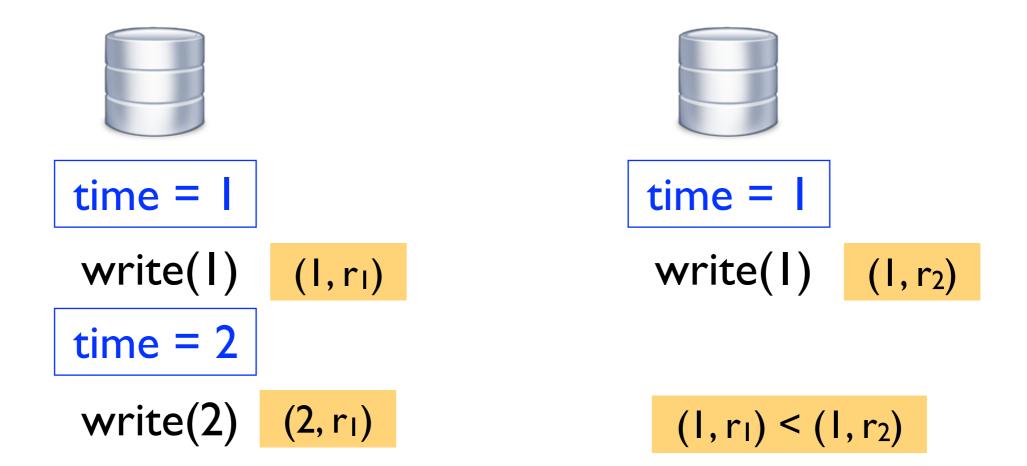
Timestamps need to be unique: ts = (CounterValue, ReplicaID)

Replica maintains a counter, incremented on each operation:



Timestamps need to be unique: ts = (CounterValue, ReplicaID) (c₁, r₁) < (c₂, r₂) \iff c₁ < c₂ \lor (c₁ = c₂ \land r₁ < r₂)

Replica maintains a counter, incremented on each operation:

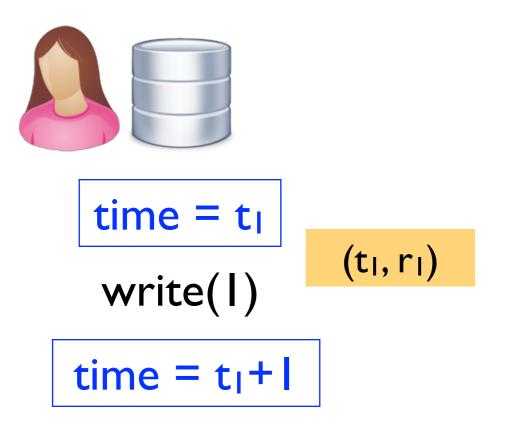


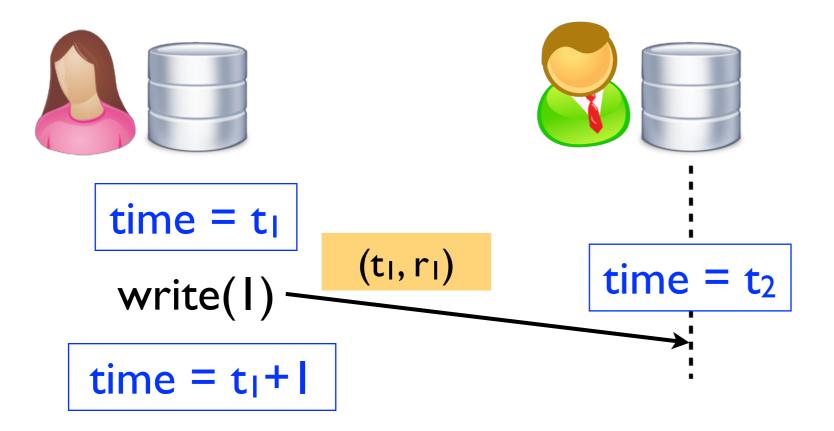
Timestamps need to be unique: ts = (CounterValue, ReplicaID) (c₁, r₁) < (c₂, r₂) \iff c₁ < c₂ \lor (c₁ = c₂ \land r₁ < r₂)

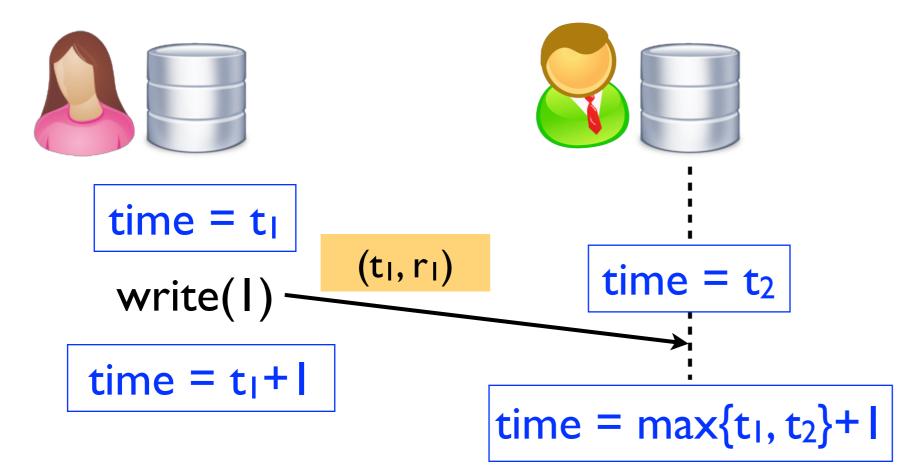


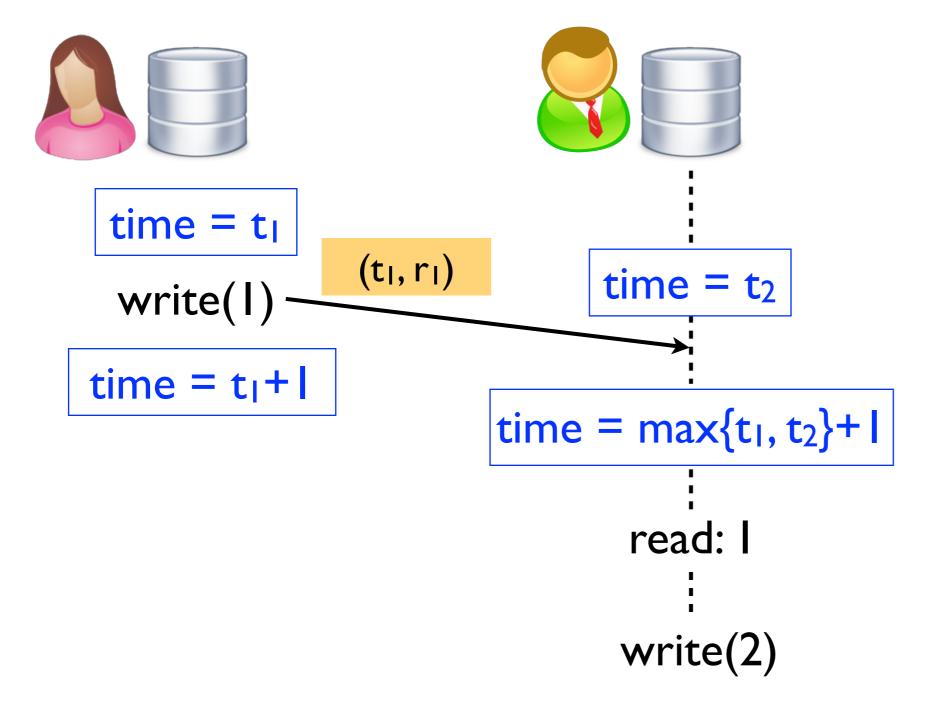


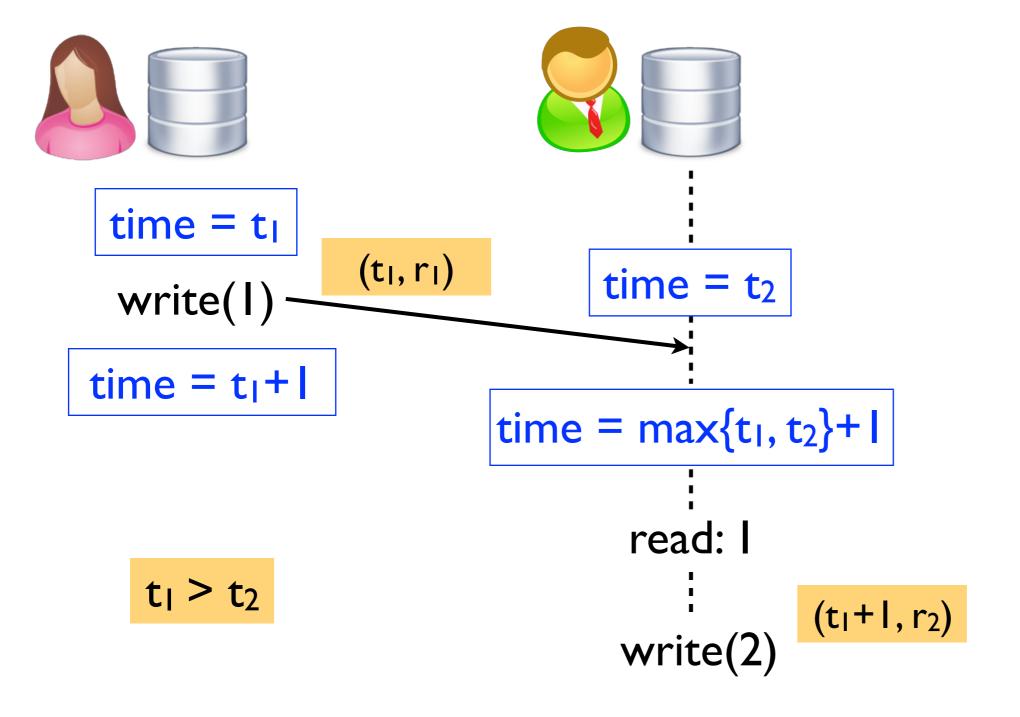
write(1)

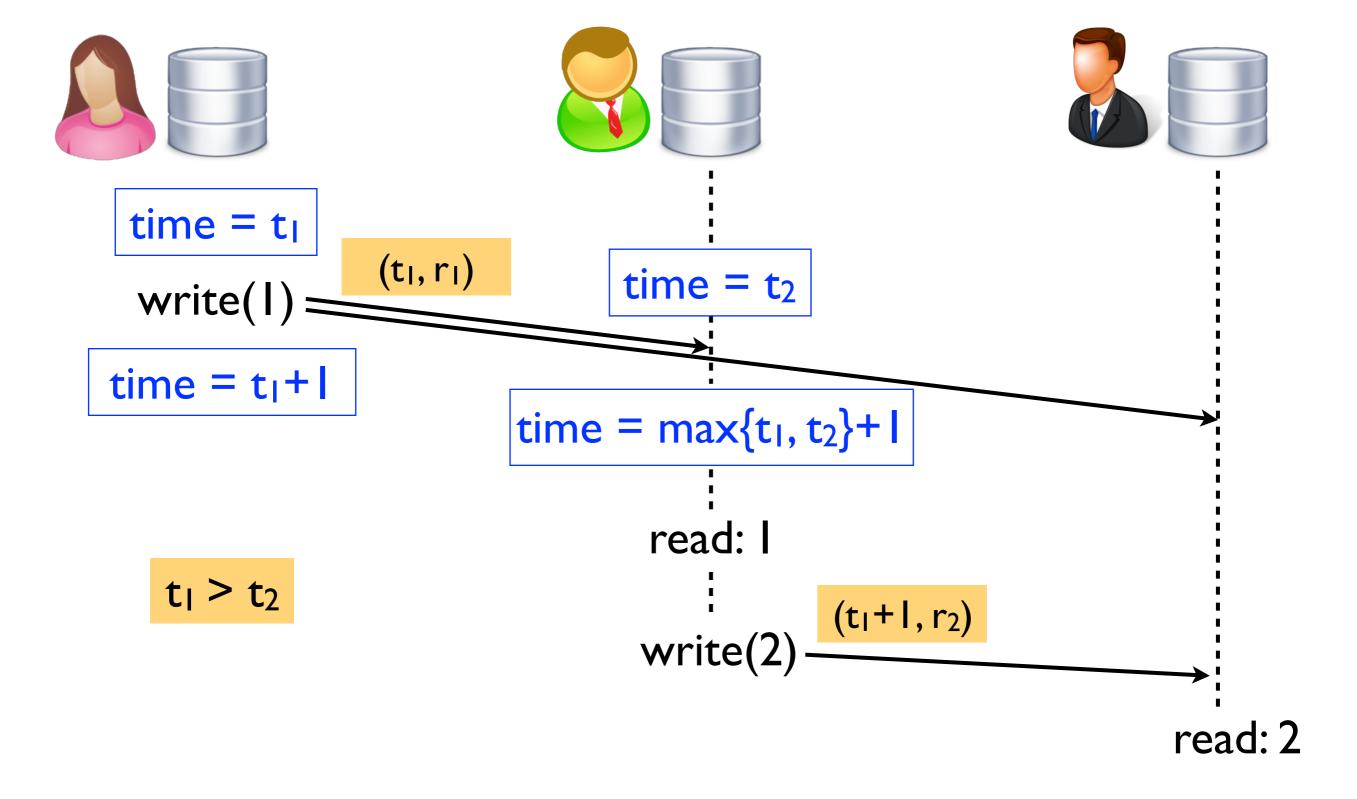


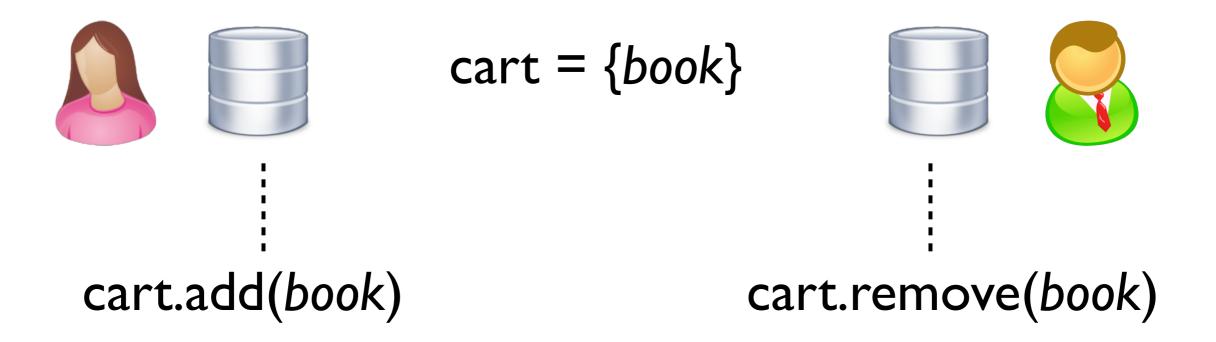


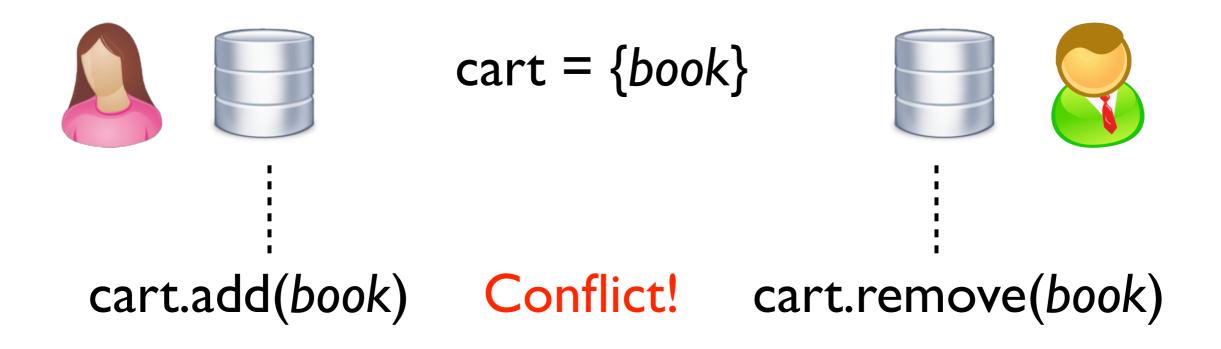


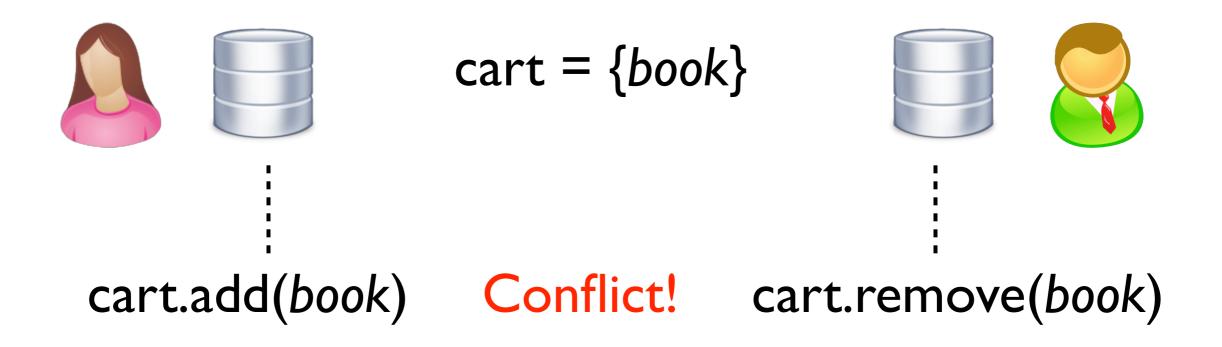




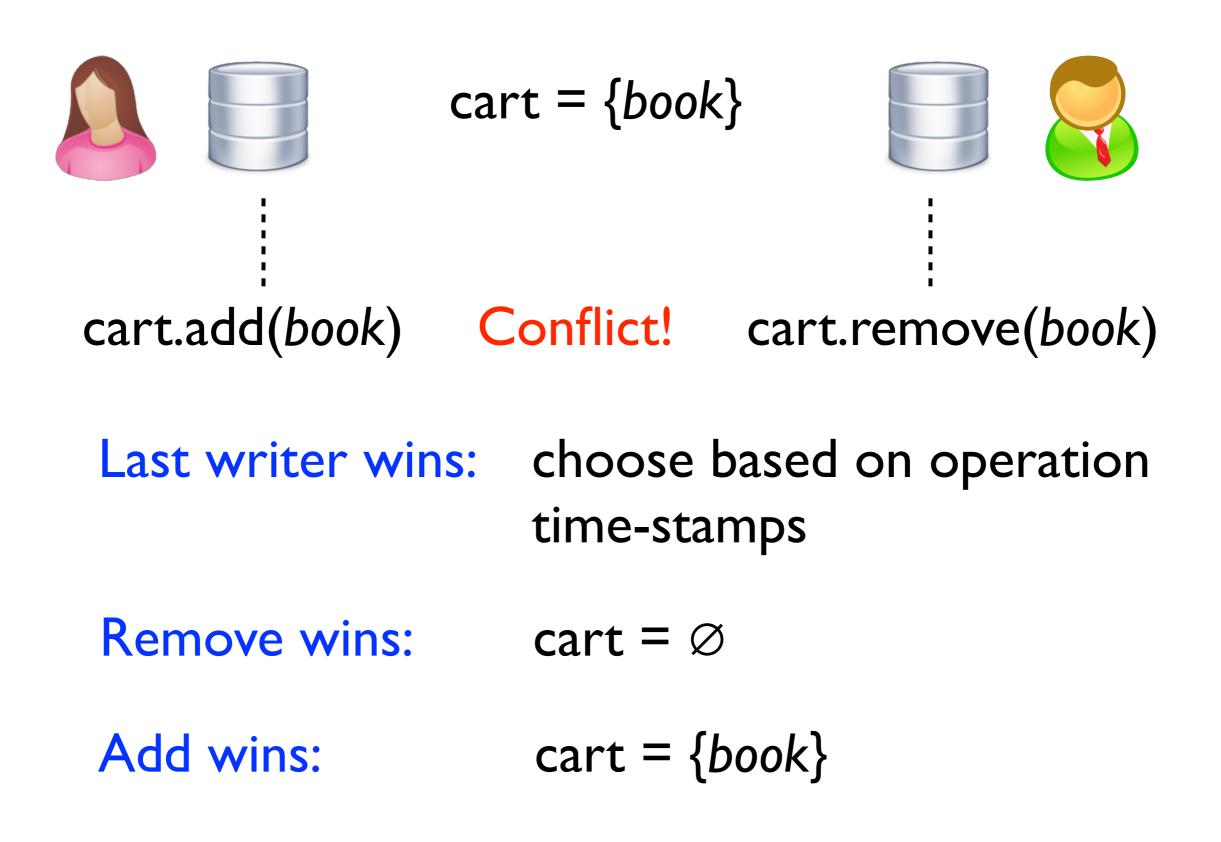




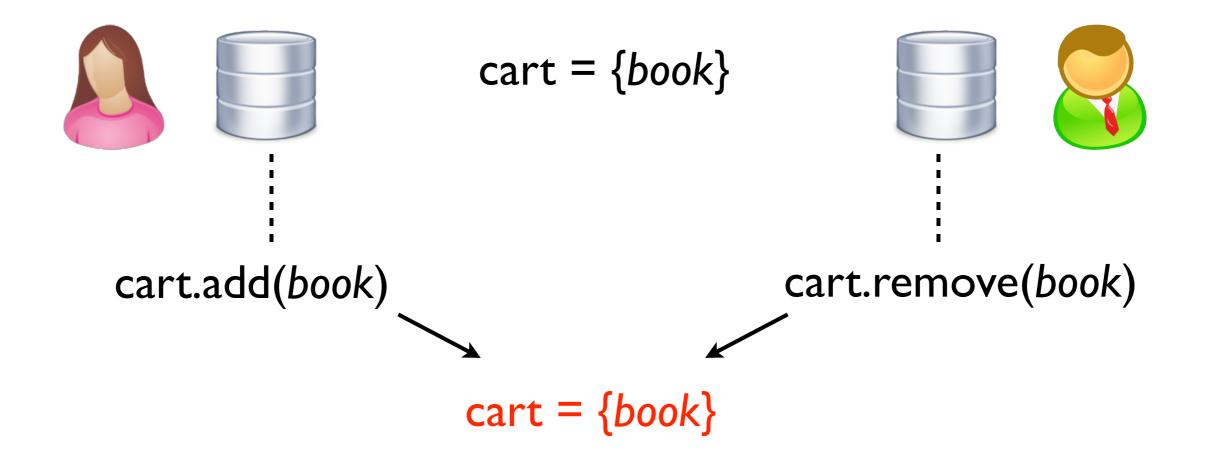




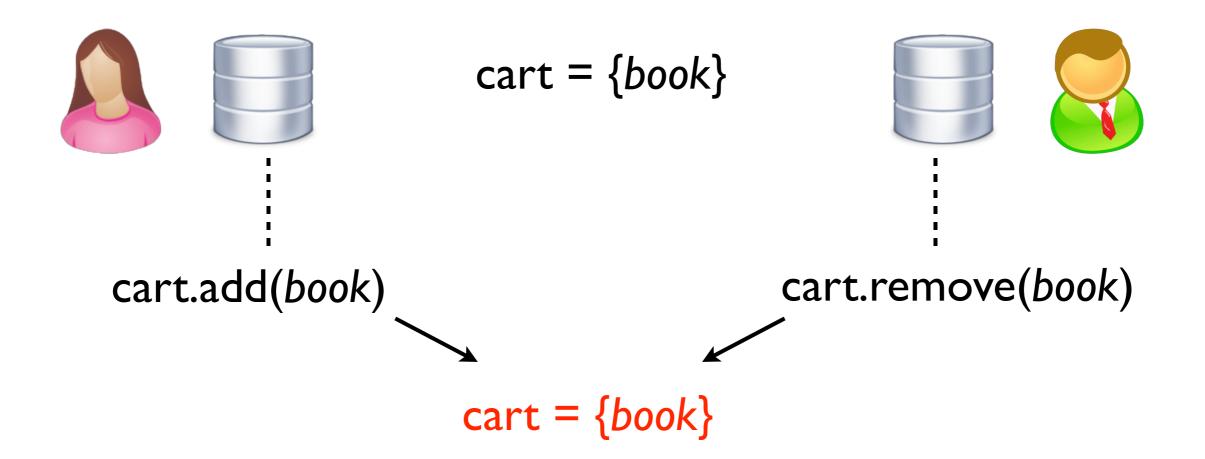
Should the remove cancel the concurrent add? Depends on application requirements



Add-wins set



Add-wins set

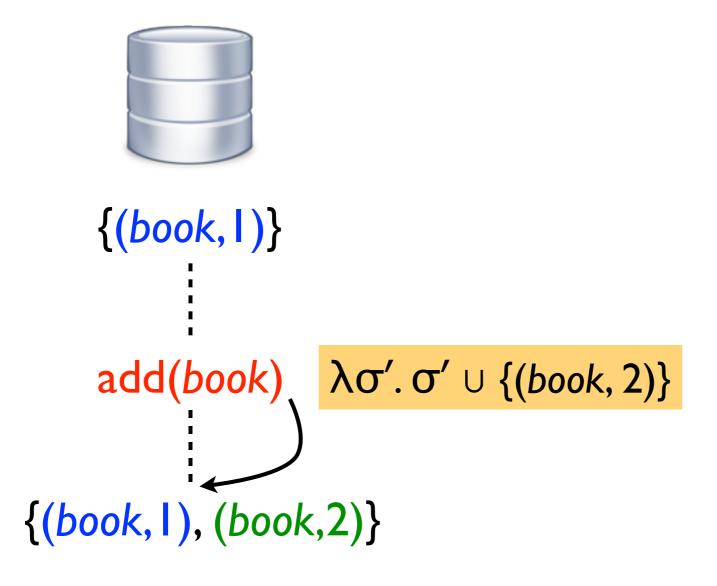


- remove() acts differently wrt add() depending on whether it's concurrent or not
- Each addition creates a new instance:
 State = set of pairs (element, unique id)

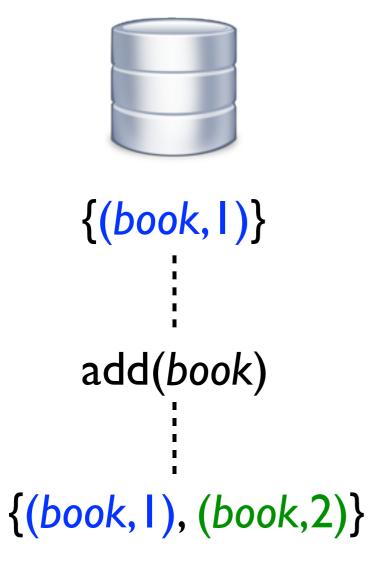


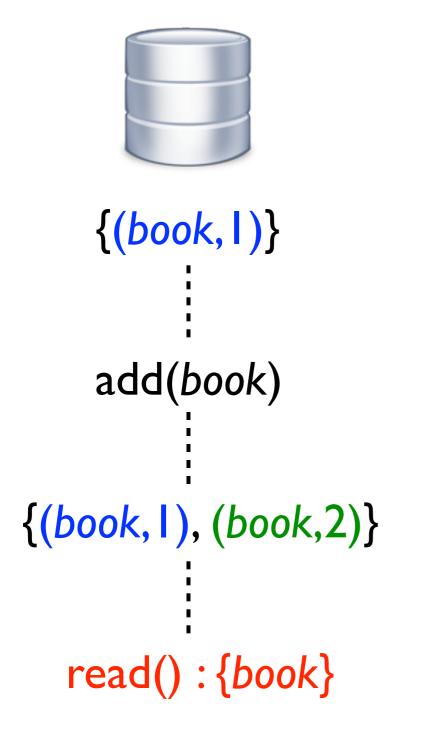
{(book, I)}

Each add() creates a new element instance: $[add(v)]_{eff}(\sigma) = \lambda \sigma' . (\sigma' \cup \{(v, uniqueid()\})$

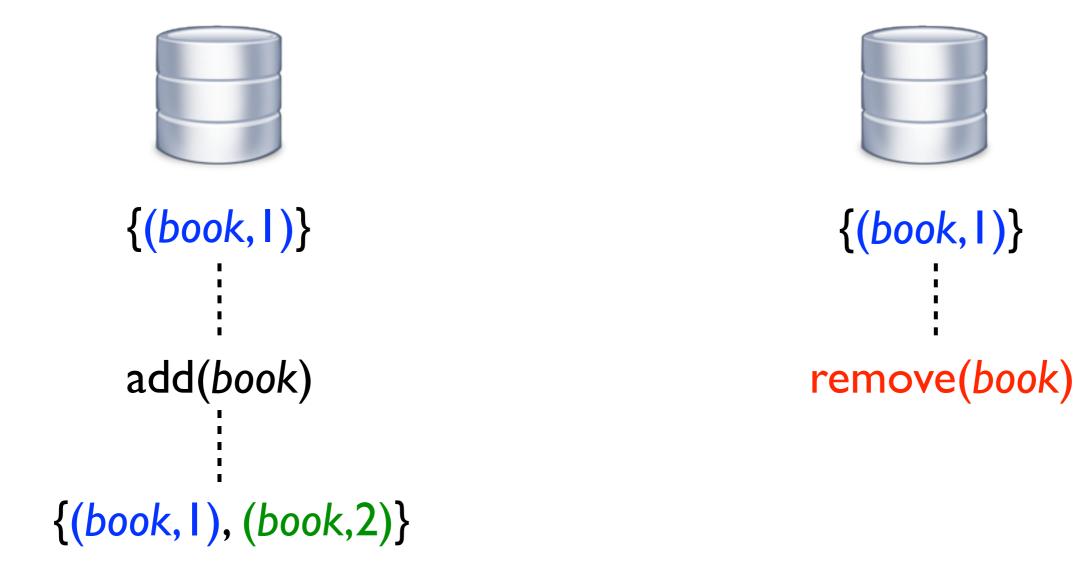


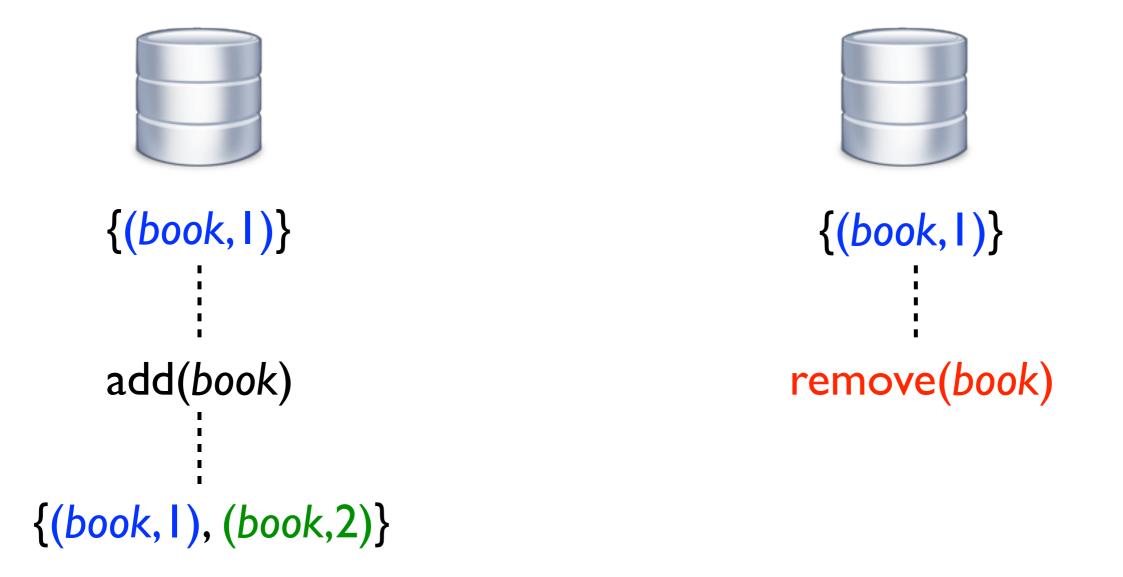
Each add() creates a new element instance: $[add(v)]_{eff}(\sigma) = \lambda \sigma' . (\sigma' \cup \{(v, uniqueid()\})$



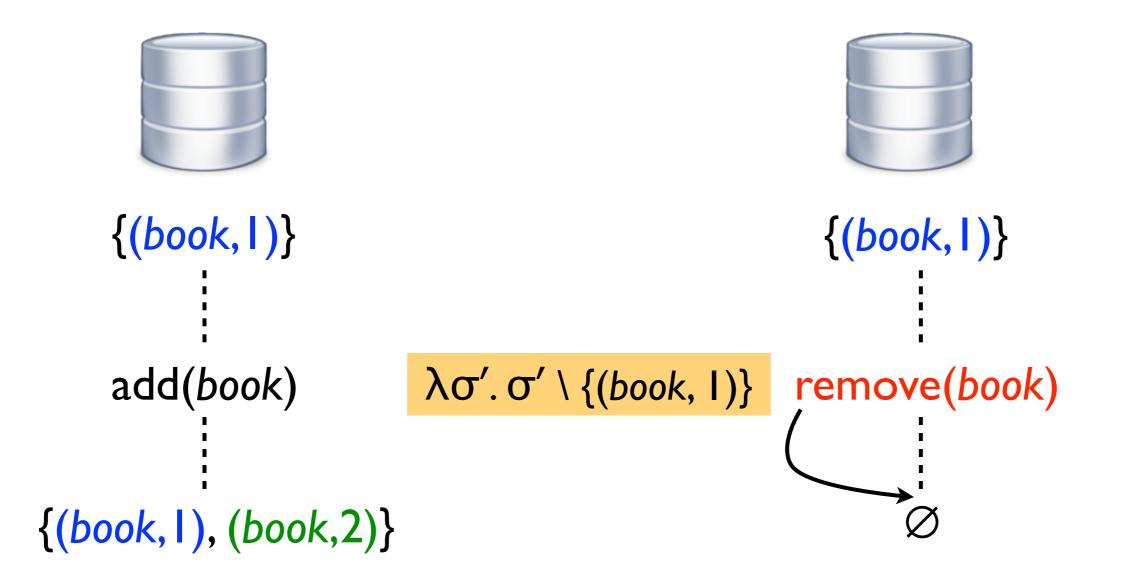


Instance ids ignored when reading the set: $[read()]_{val}(\sigma) = \{v \mid \{\exists id. (v, id)\} \in \sigma\}$

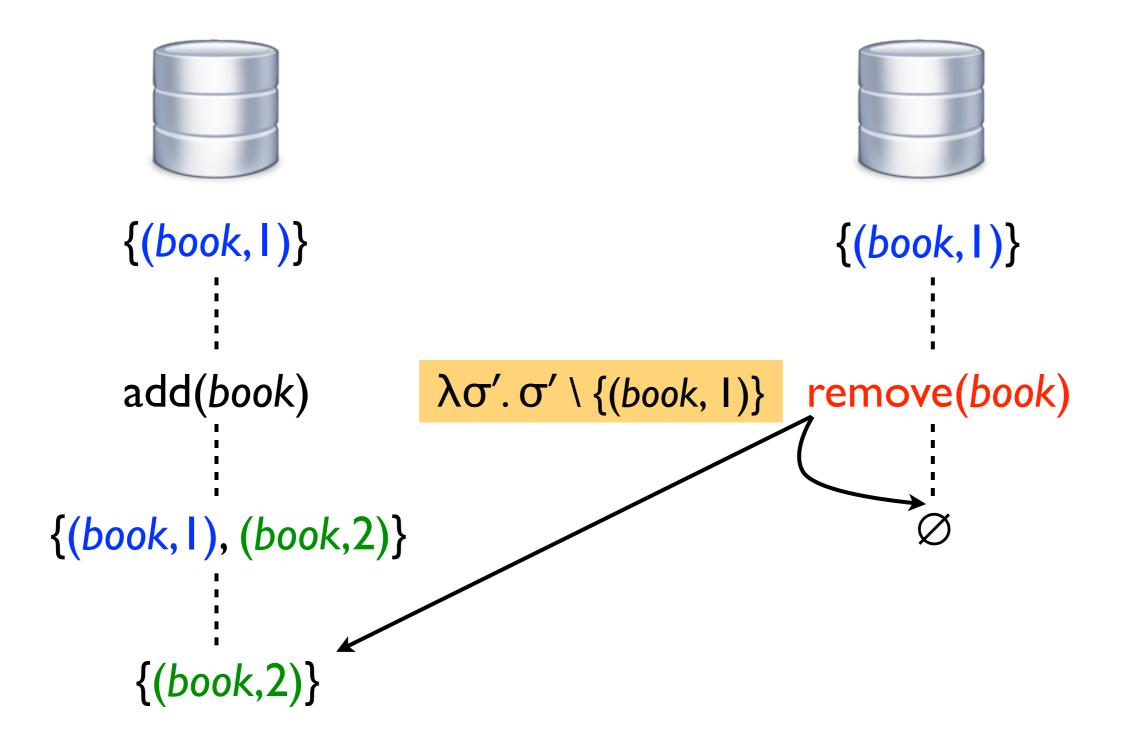




remove(v) removes all currently present instances of x: $[remove(v)]_{eff}(\sigma) = \lambda \sigma' . (\sigma' \setminus \{(v, id) \in \sigma\})$

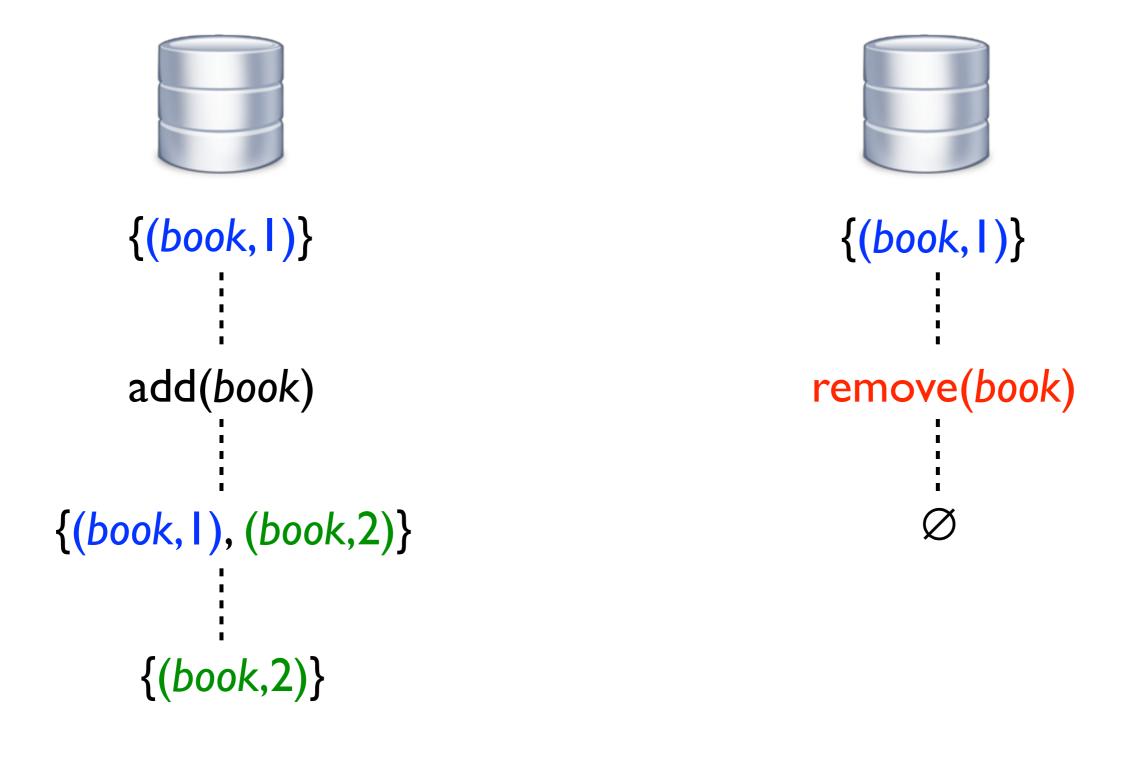


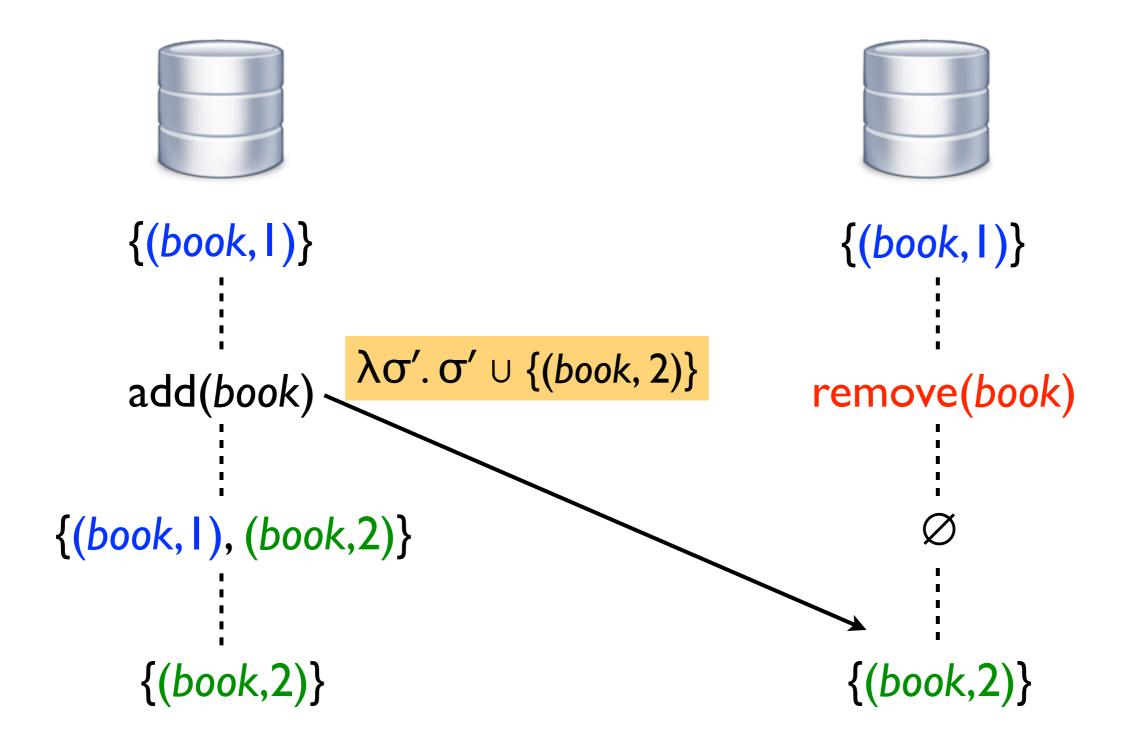
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remove(v) removes all currently present instances of x:

 $[\![remove(v)]\!]_{eff}(\sigma) = \lambda\sigma'. (\sigma' \setminus \{(v, id) \in \sigma\})$





Effectors commutative \rightarrow replicas converge

Take-aways

• Need to ensure commutativity to guarantee quiescent consistency

Need to make choices about how to resolve conflicts

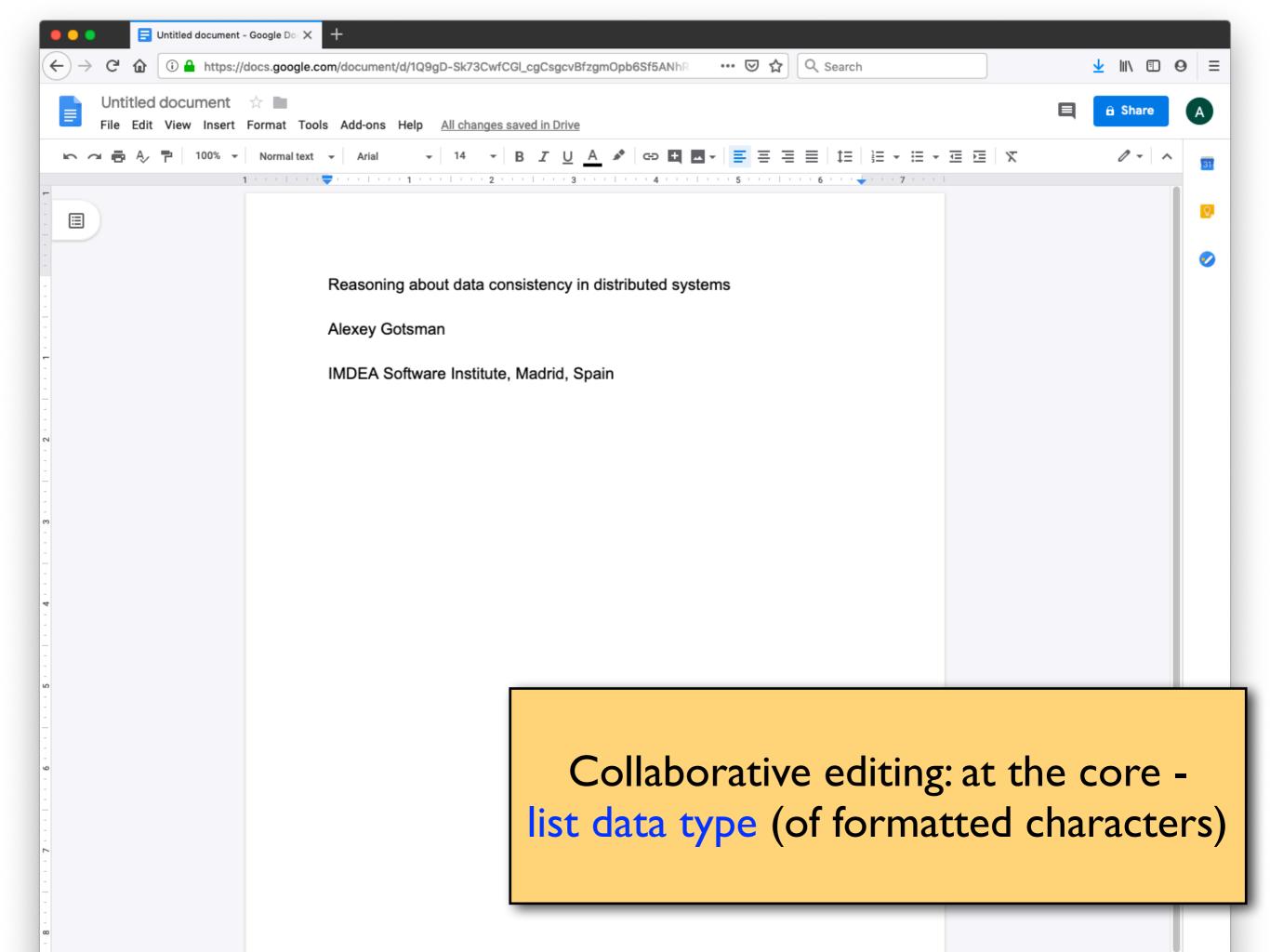
Replicated data type uses

• Provided by some data stores:



• Implemented by programmers on their own:





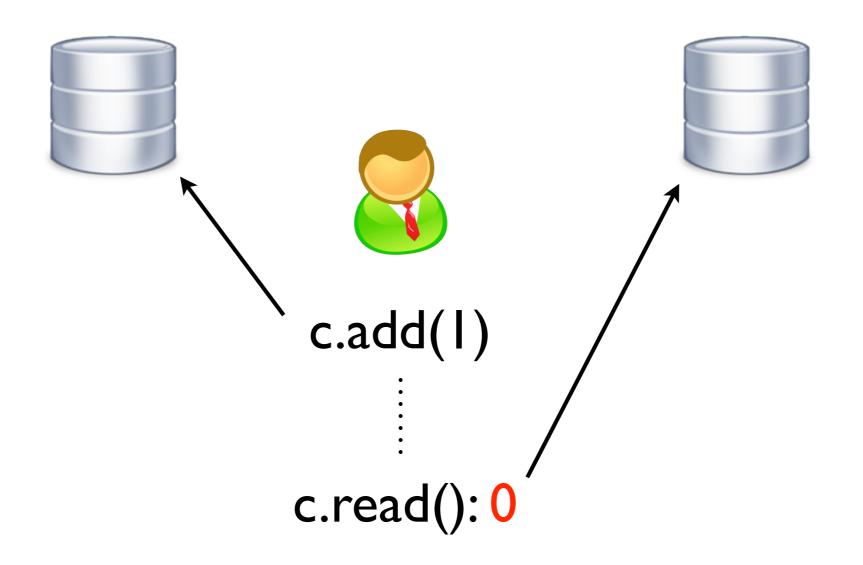
Operational specification

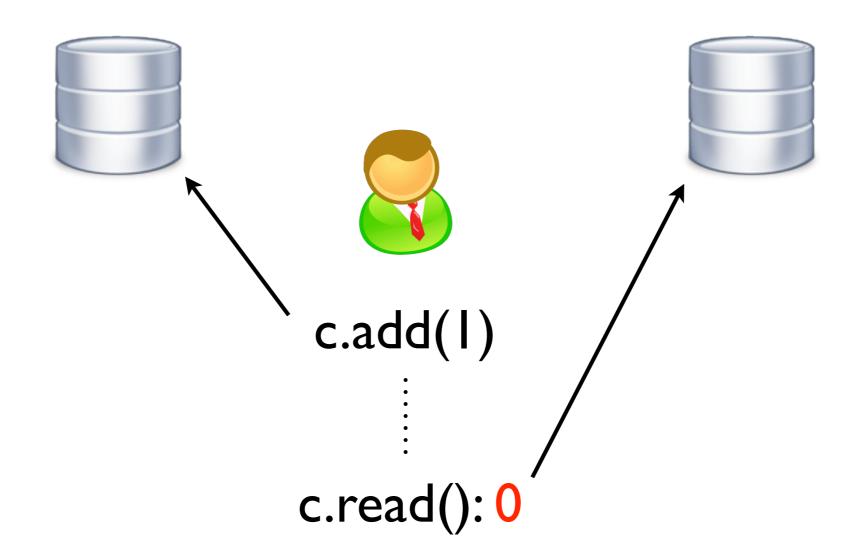
- Given a database with a set of objects of replicated data types
- Eventual consistency model = set of all histories produced by arbitrary client interactions with the data type implementations (with any allowed message deliveries)
- Implies quiescent consistency: if no new updates are made to the database, then replicas will eventually converge to the same state

Eventual consistency and replicated data types, axiomatically



c.add(1)





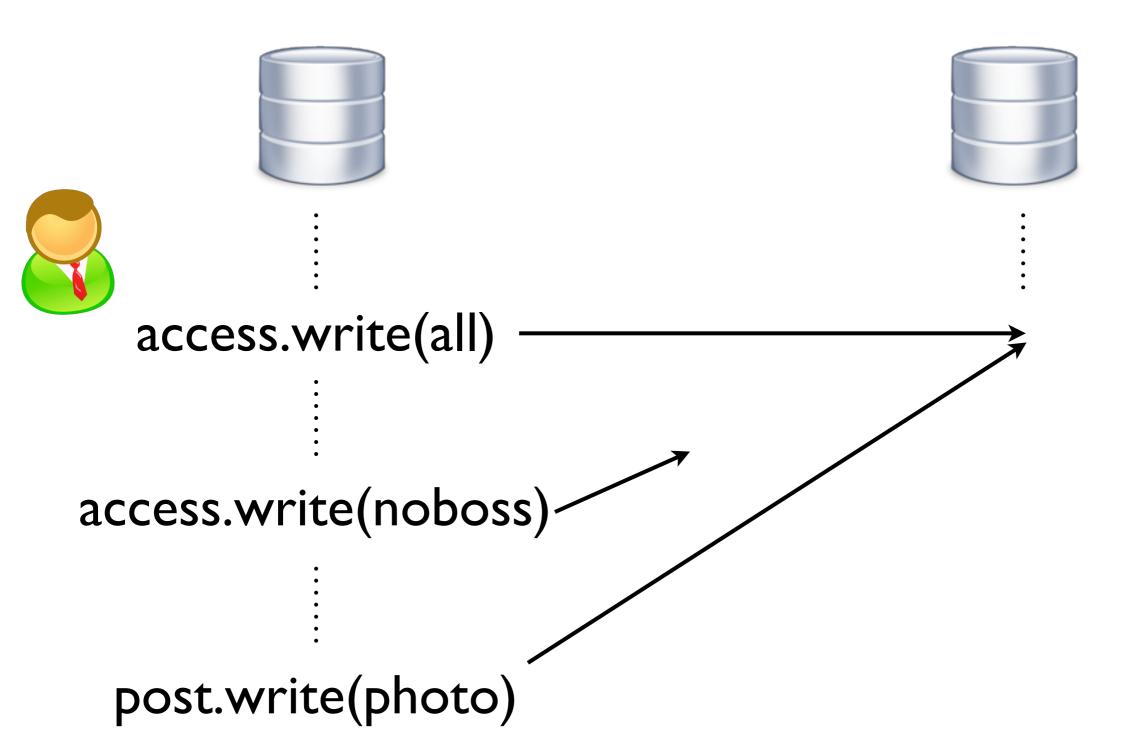
Can be disallowed if the client sticks to the same replica: Read Your Writes guarantee

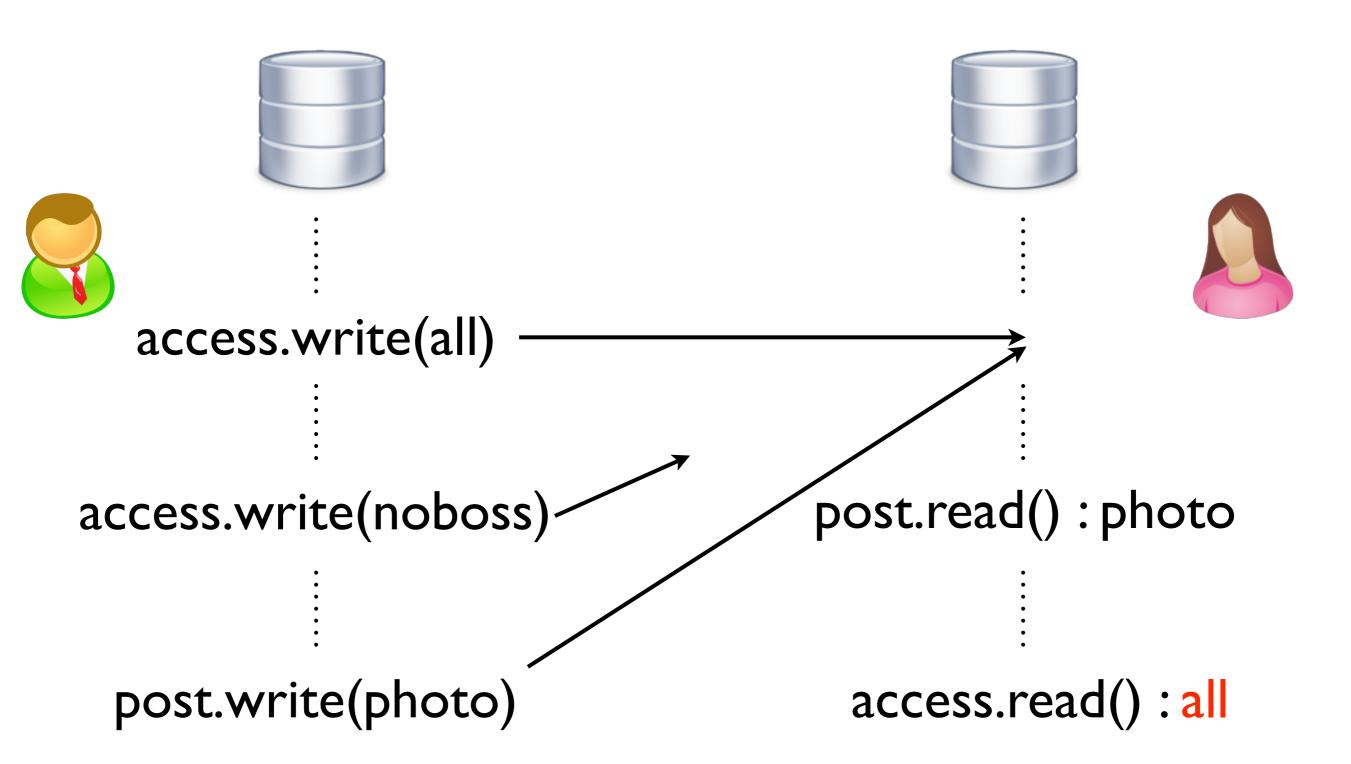


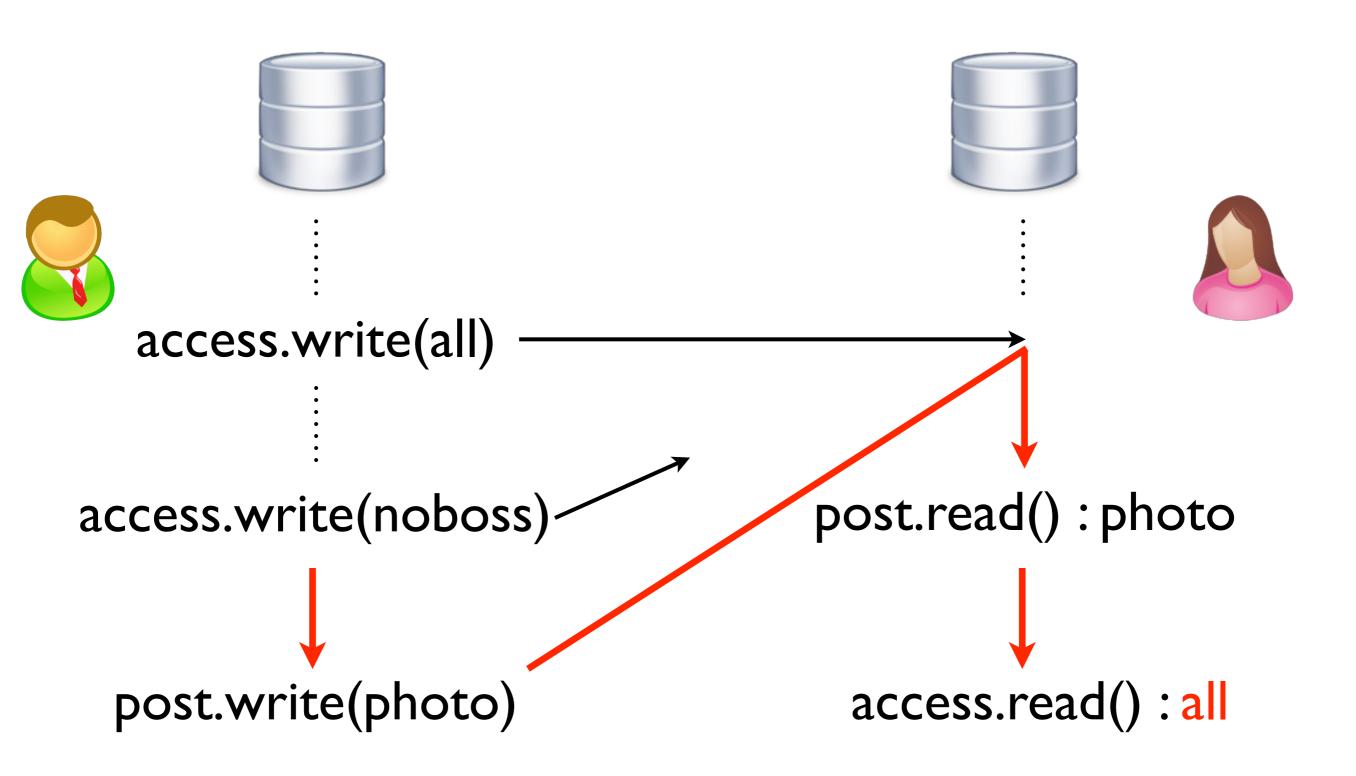


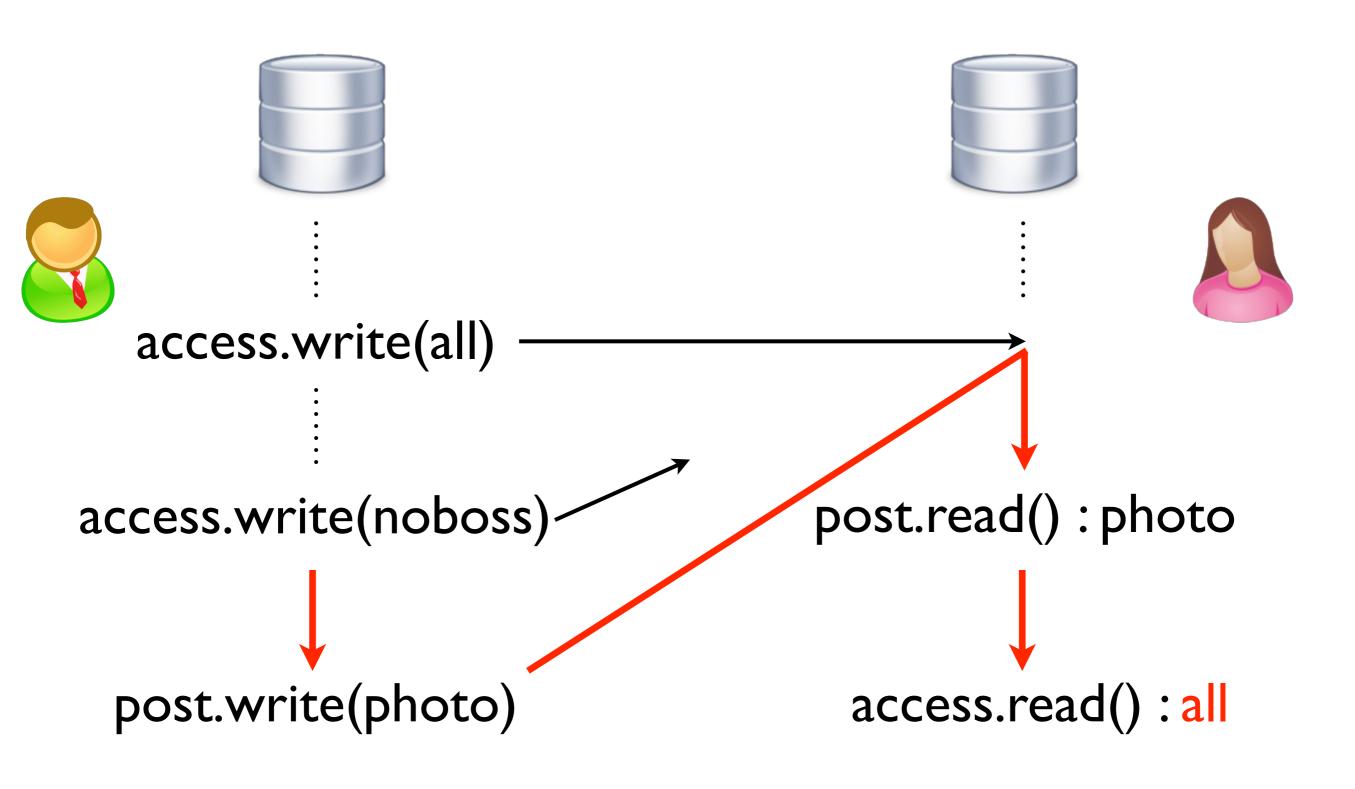
access.write(noboss)

post.write(photo)

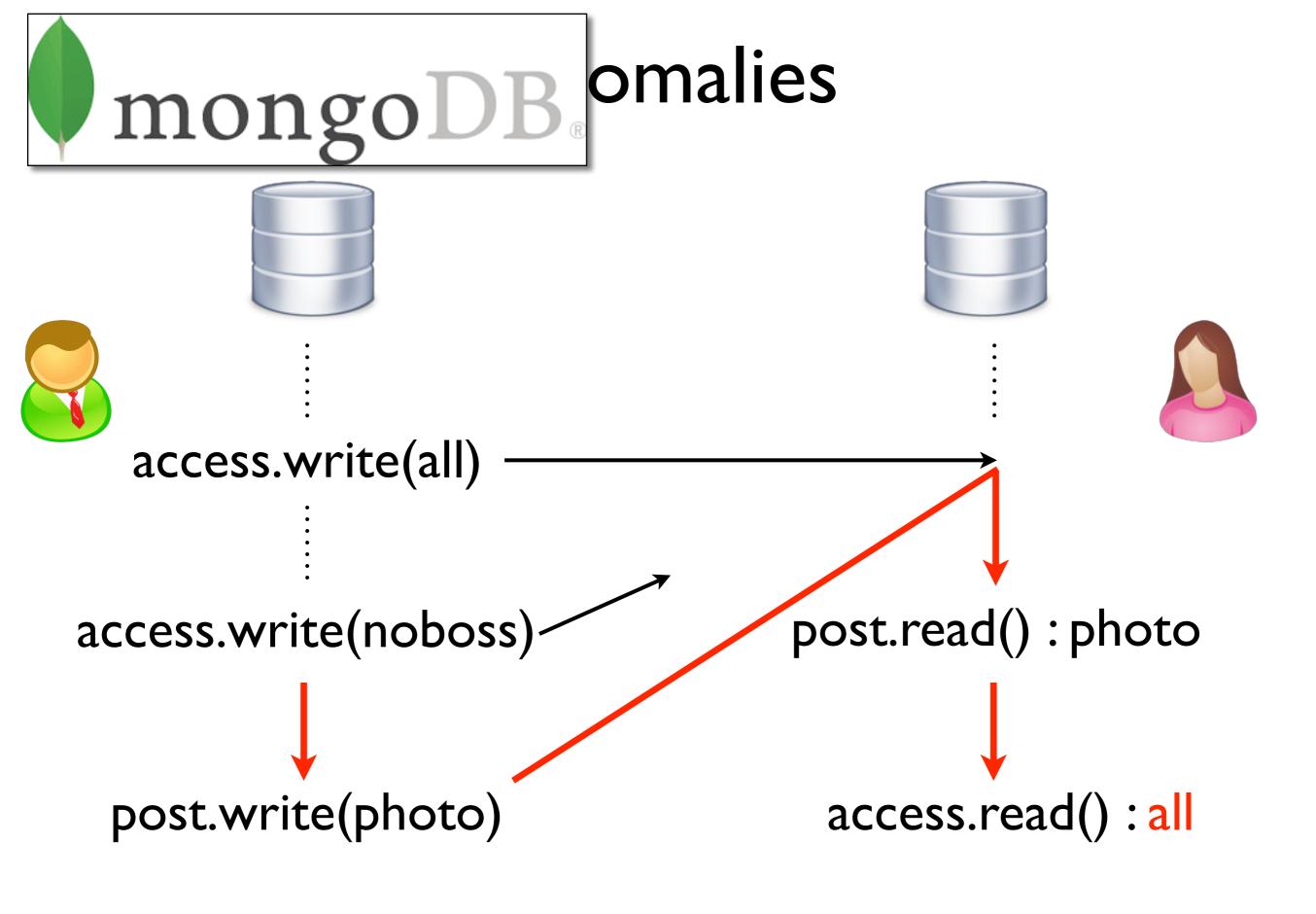








Causality violation: disallowed by causal consistency



Causality violation: disallowed by causal consistency

Specification

- Lots of replicated data type implementations: e.g., can send snapshots of object states instead of operations
- Lots of message delivery guarantees: different implementations of causal consistency
- Want specifications that abstract from implementation details: both replicated data types and anomalies

Axiomatic specifications

- Choose a set of relations over events: r₁, ..., r_n
 Abstractly specify essential information about how operations are processed inside the system
- Abstract execution (H, r₁, ..., r_n)
- Choose a set of axioms *A* constraining abstract executions
- Consistency model = {H | $\exists r_1, ..., r_n$. (H, $r_1, ..., r_n$) $\vDash \mathscr{A}$ }

Sequential consistency

(E,so) $| \exists$ total order to. (E, so, to) satisfies:

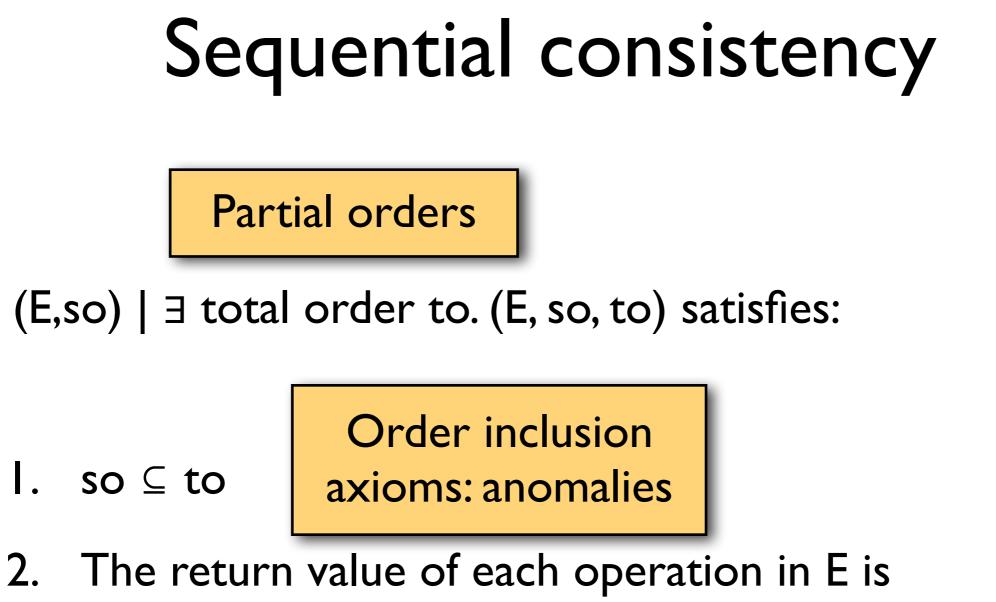
- I. so \subseteq to
- 2. The return value of each operation in E is computed from a state obtained by executing all operations on the same object preceding it in to

Sequential consistency

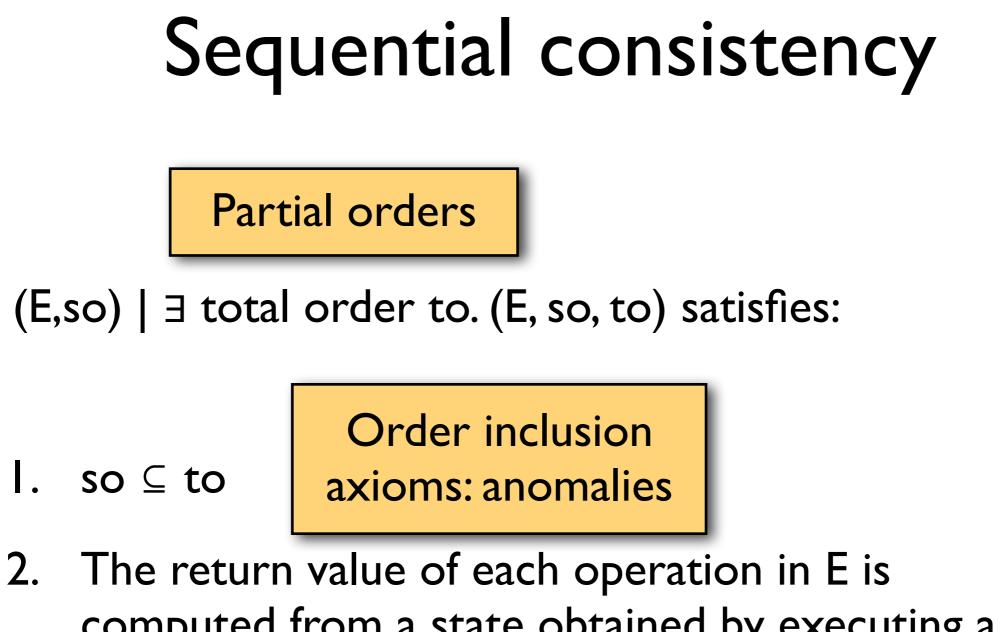
Partial orders

(E,so) | \exists total order to. (E, so, to) satisfies:

- I. so \subseteq to
- 2. The return value of each operation in E is computed from a state obtained by executing all operations on the same object preceding it in to

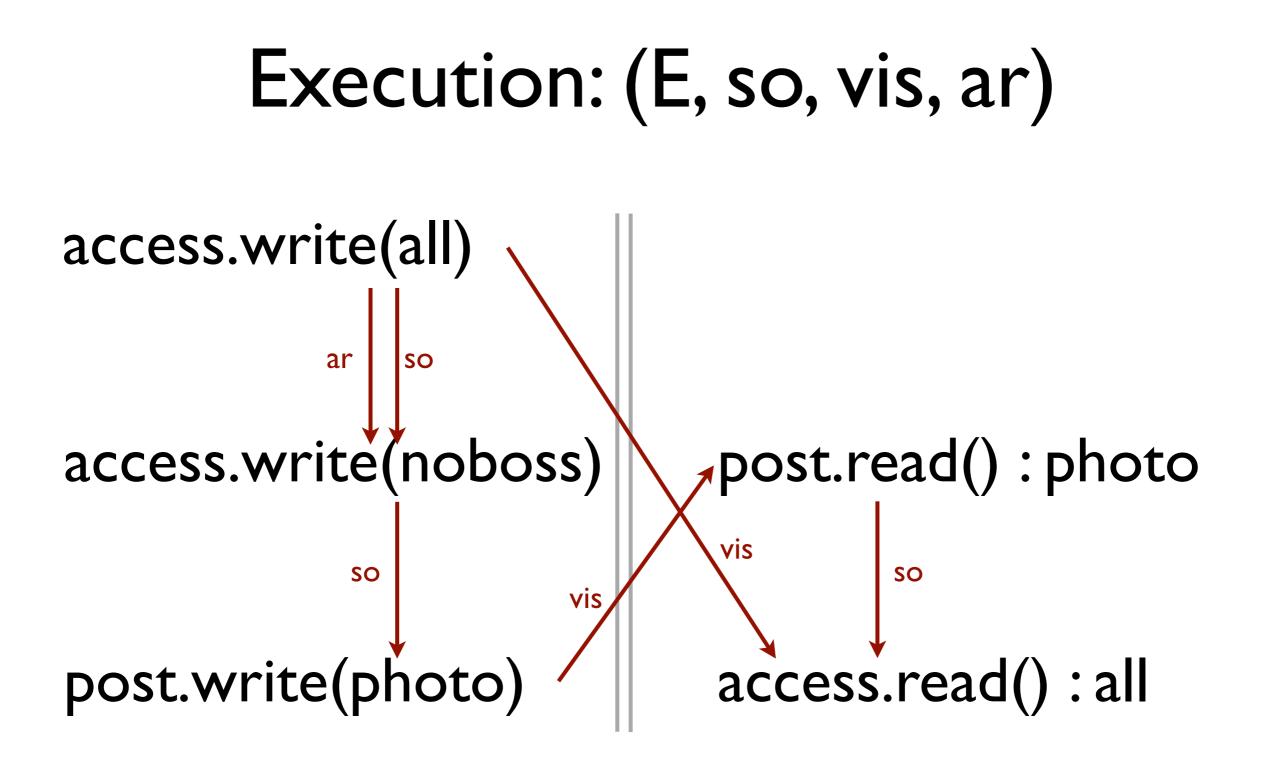


computed from a state obtained by executing all operations on the same object preceding it in to

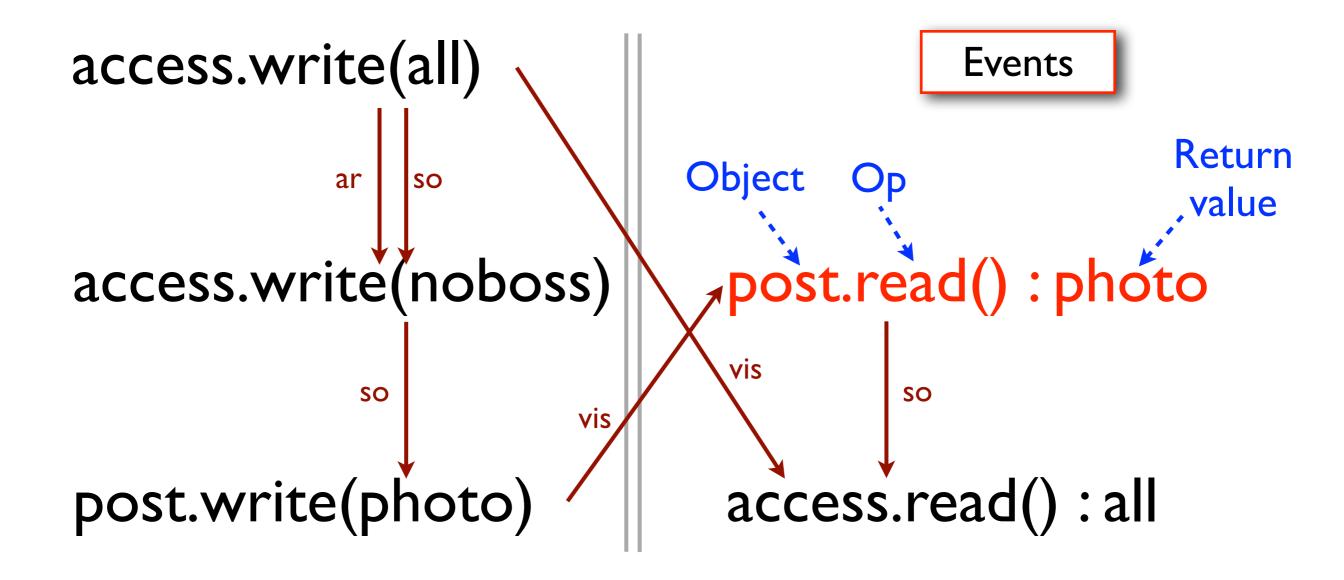


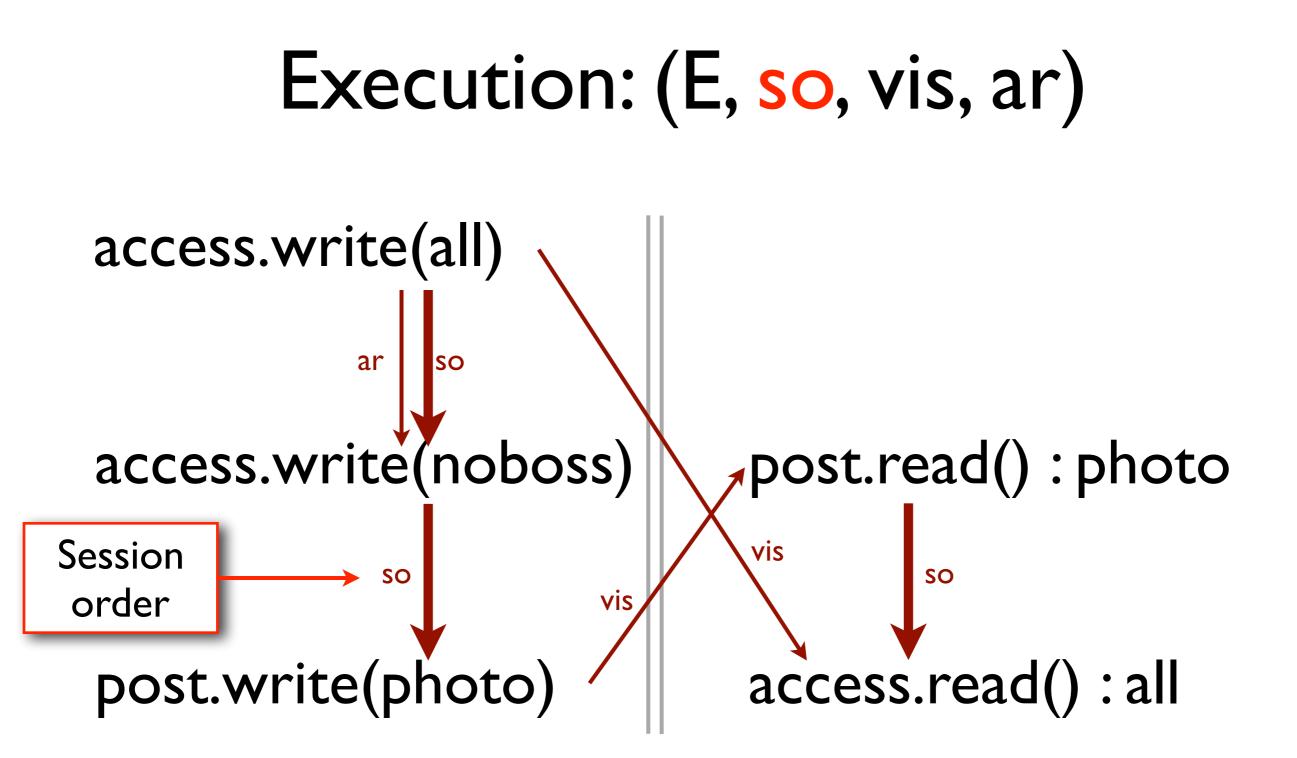
computed from a state obtained by executing all operations on the same object preceding it in to

Return value axiom: replicated data types

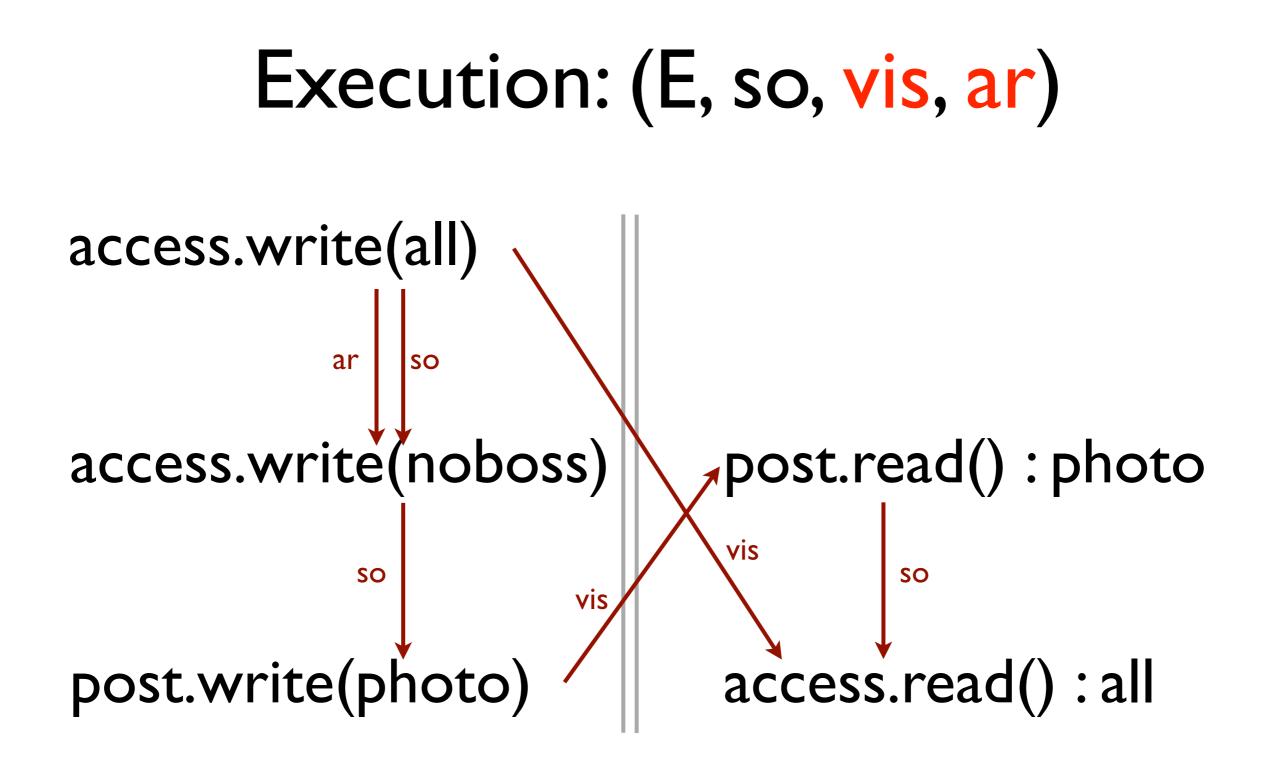


Execution: (E, so, vis, ar)

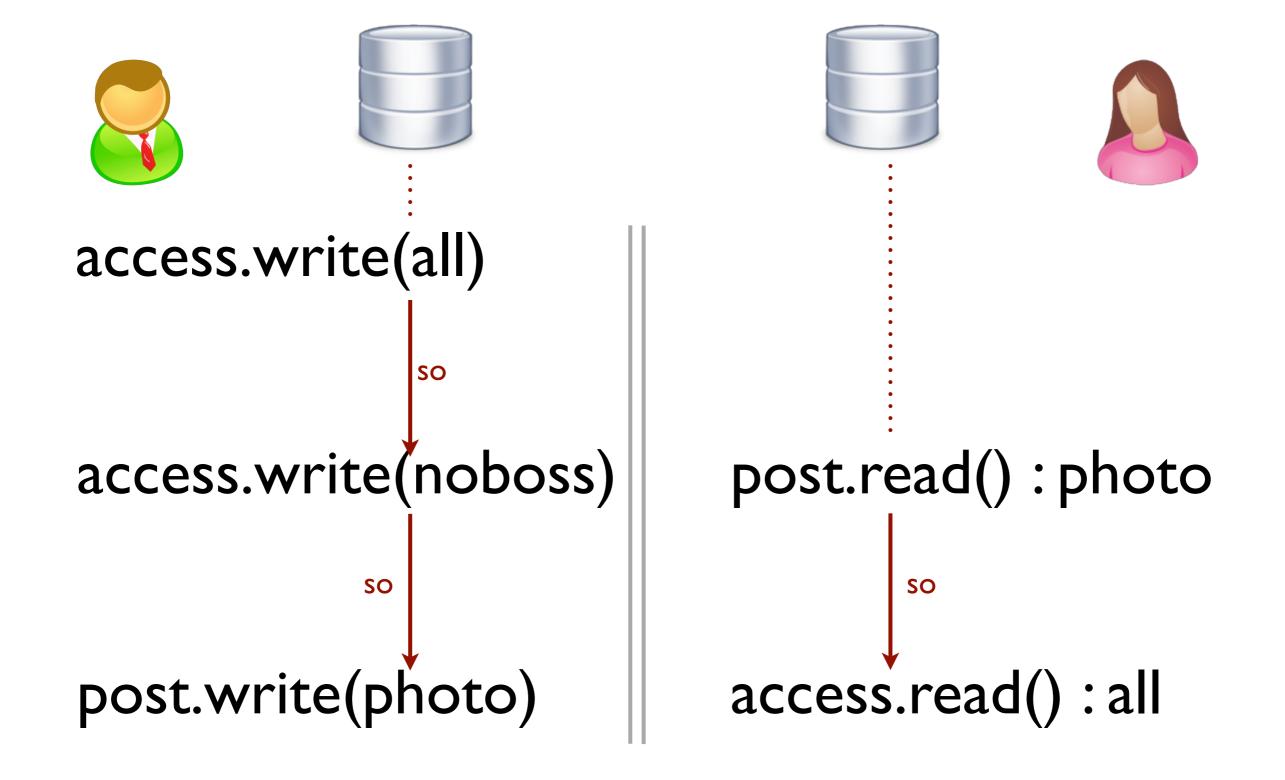


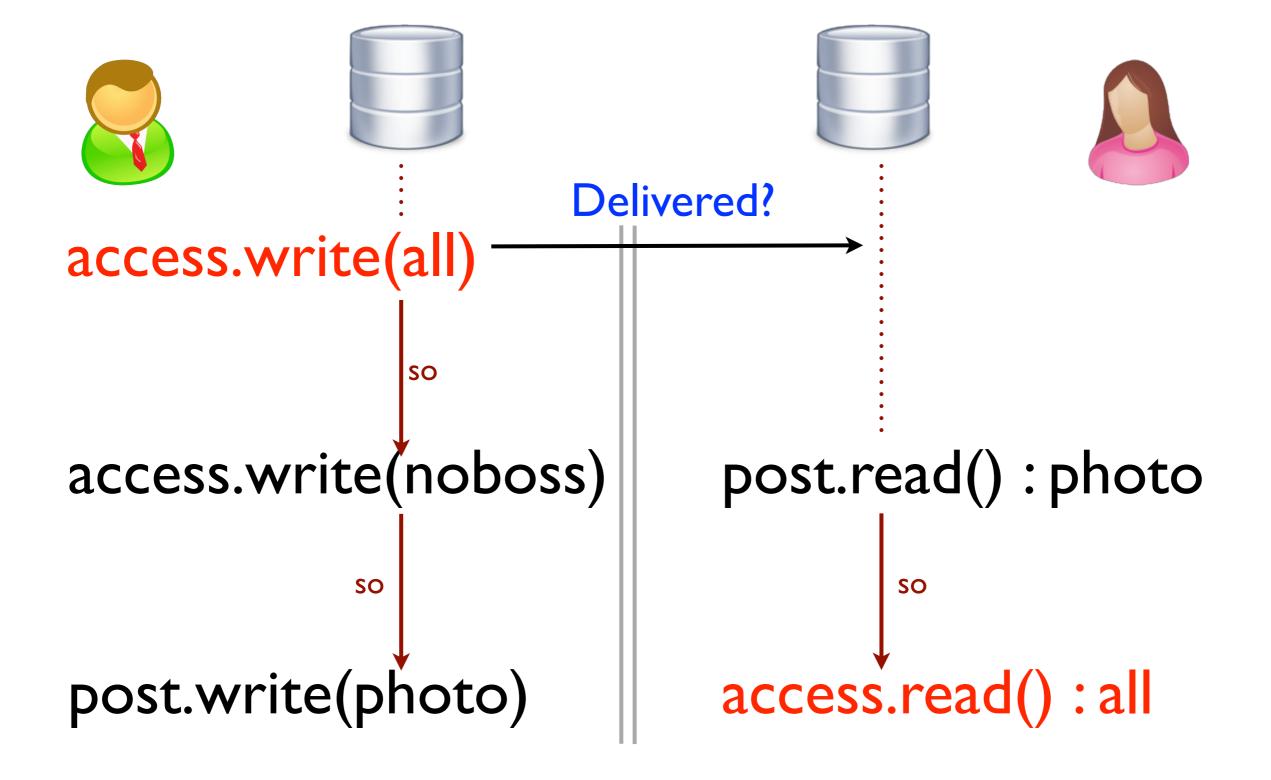


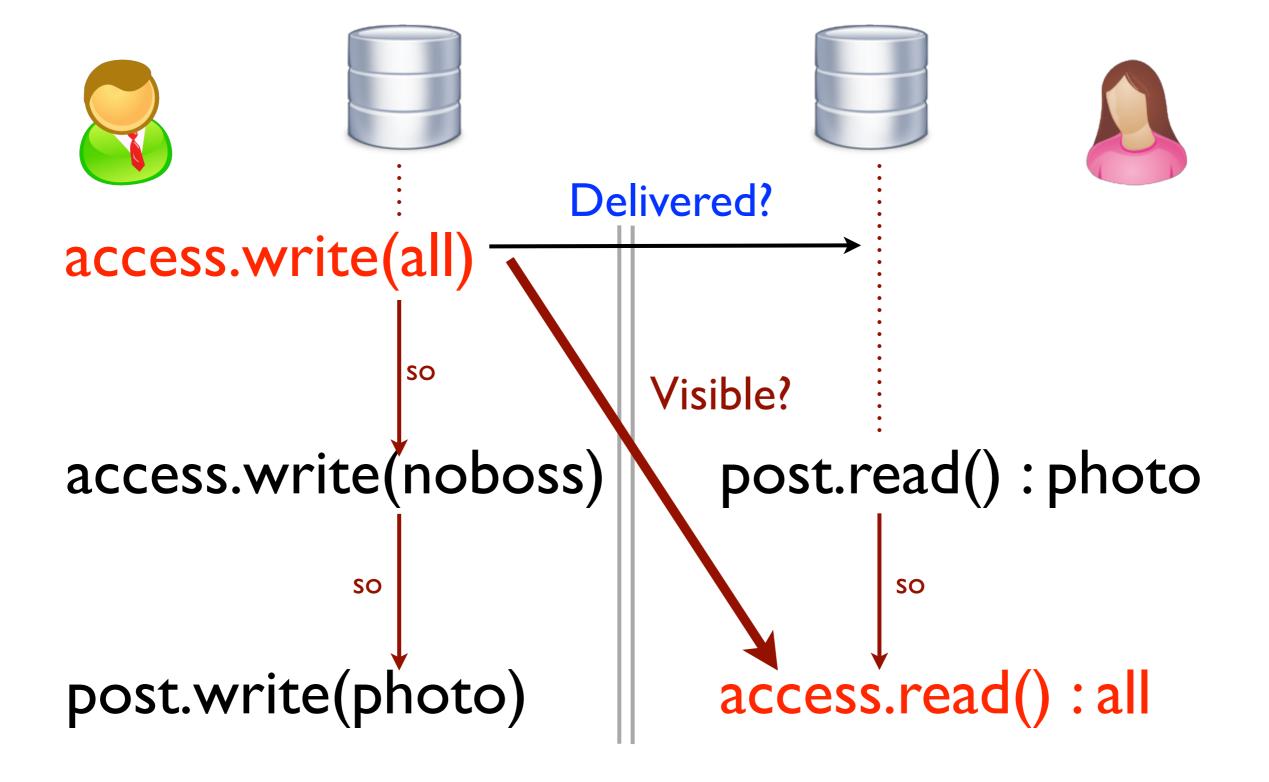
The order of requests by the same session

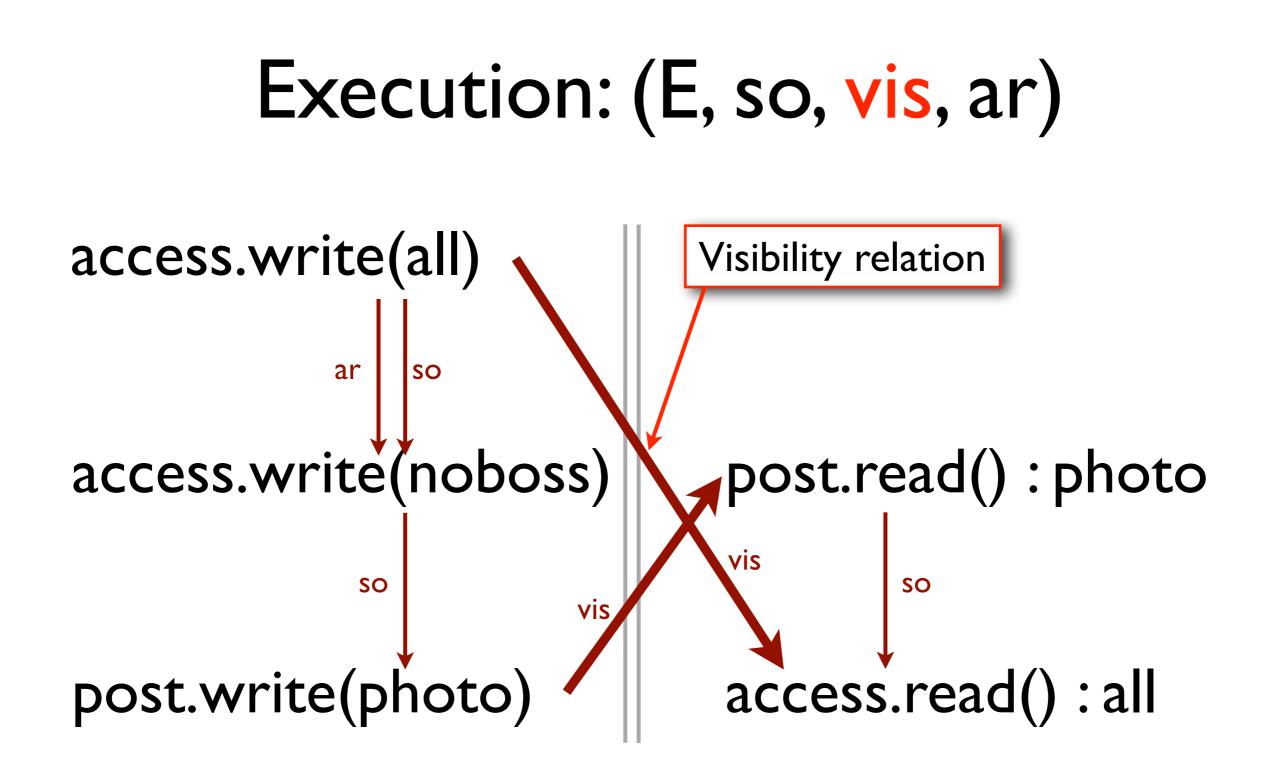


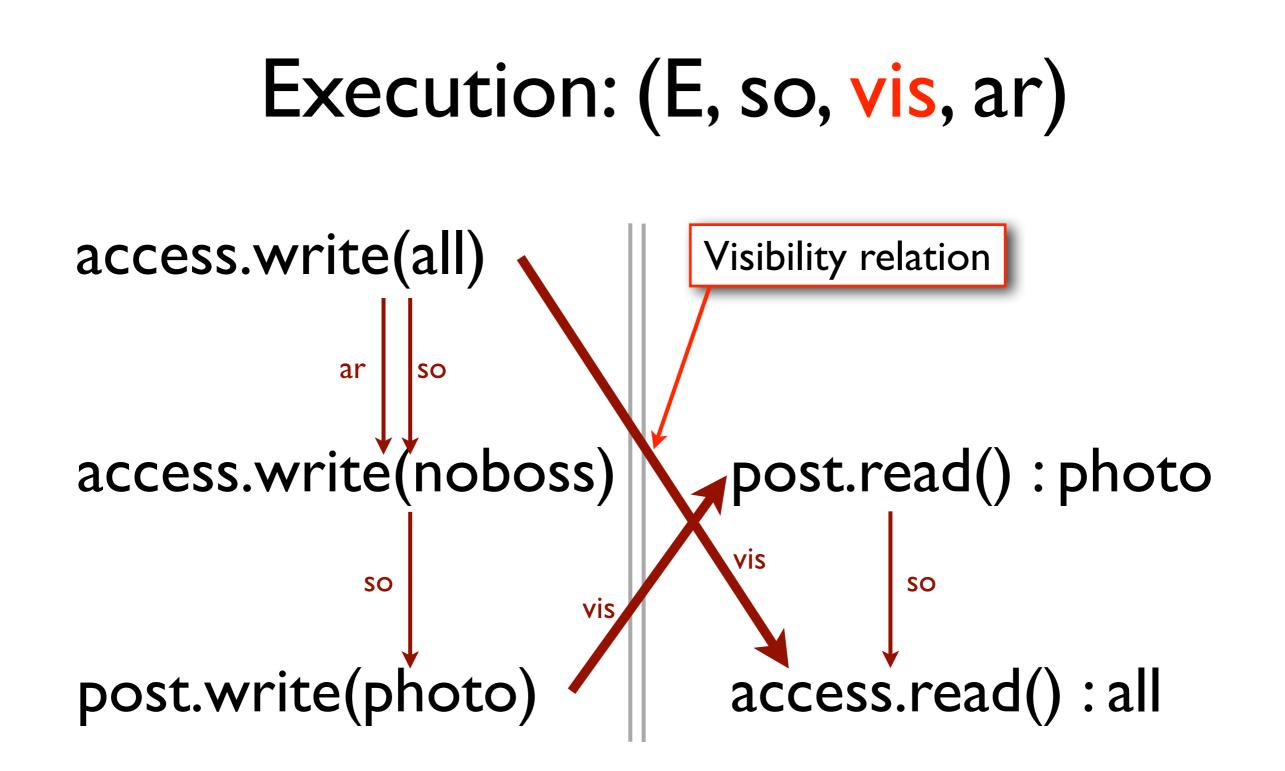
Declaratively specify ways in which the database processes requests



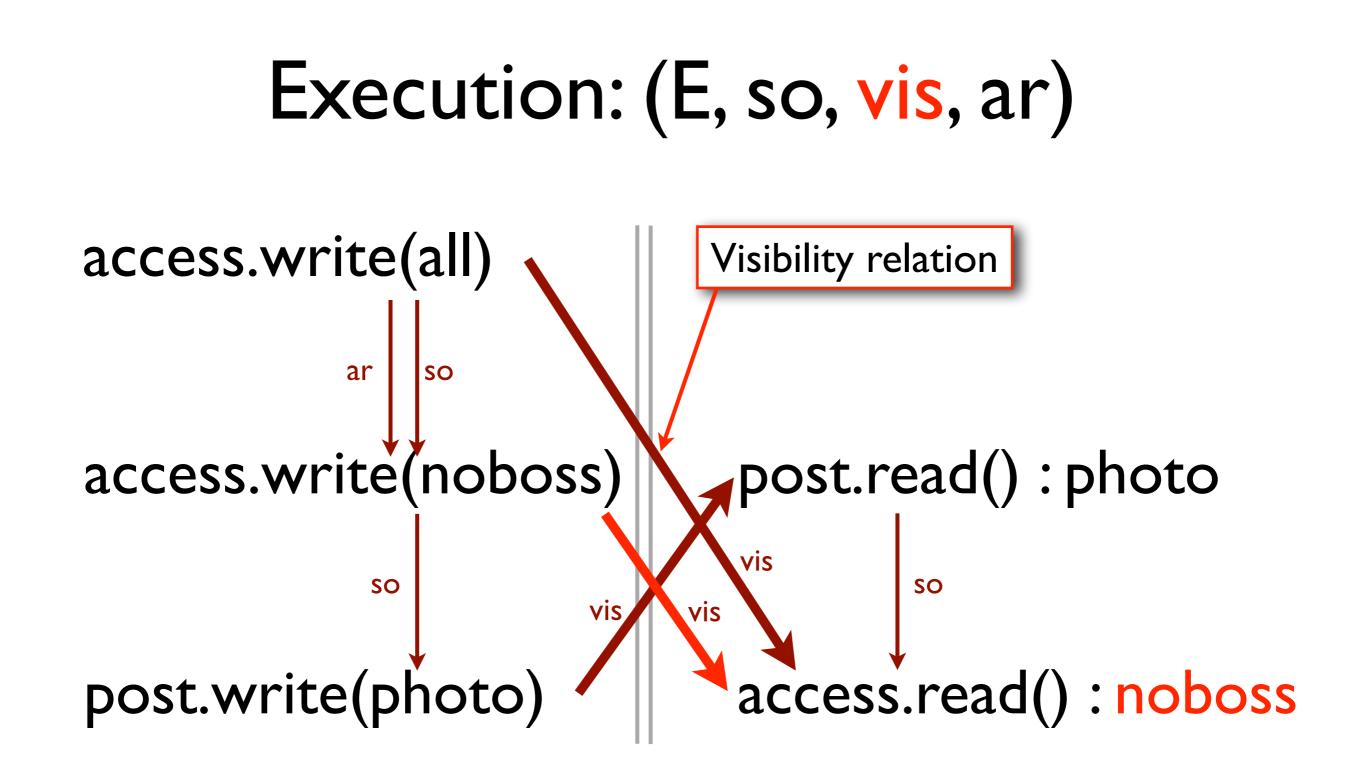




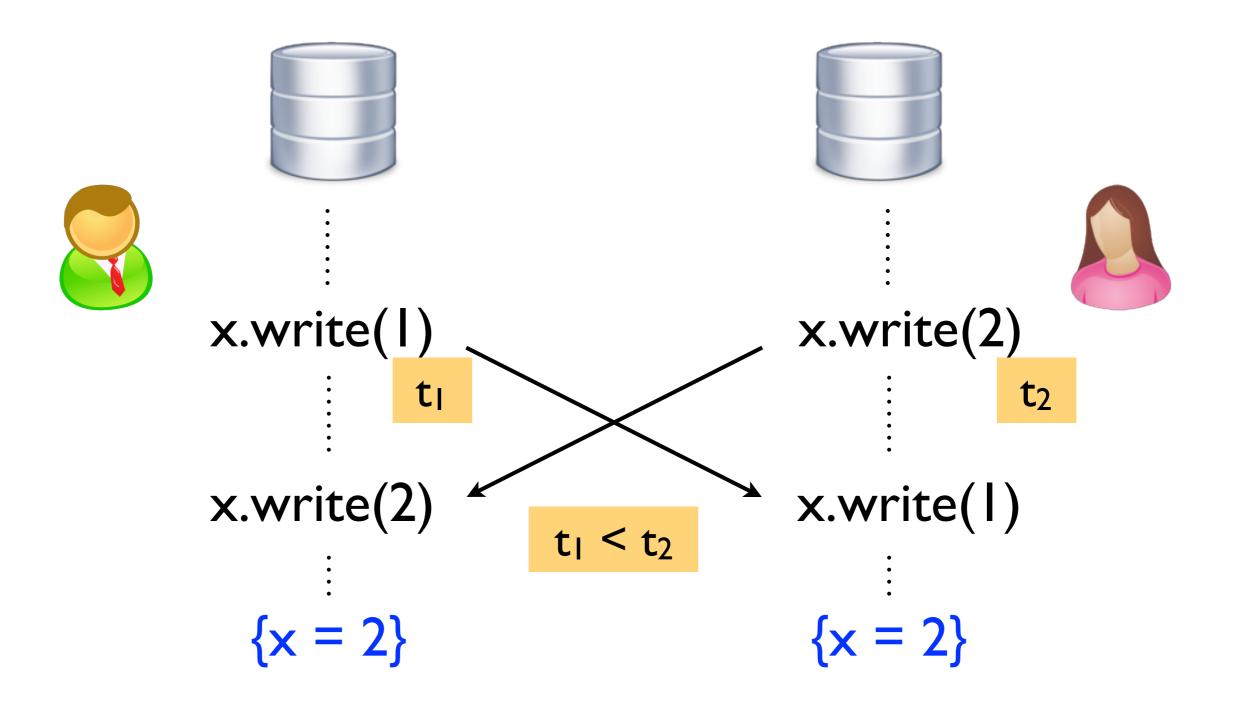




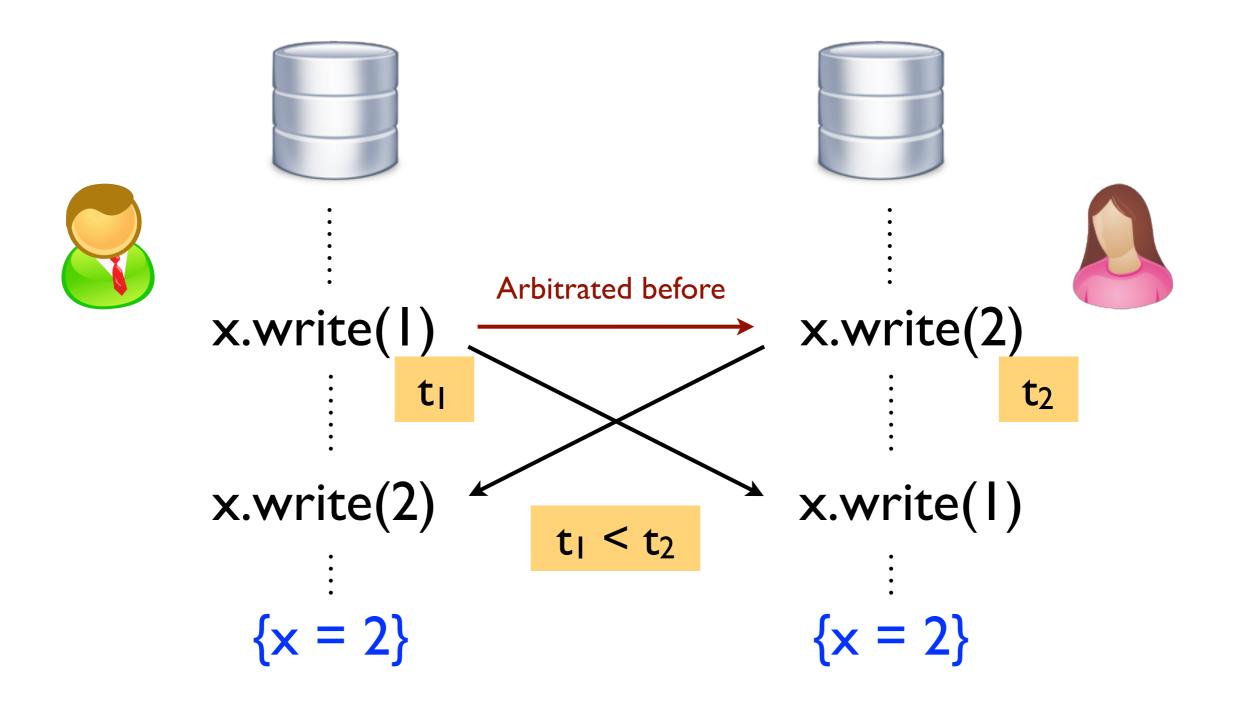
vis is irreflexive and acyclic



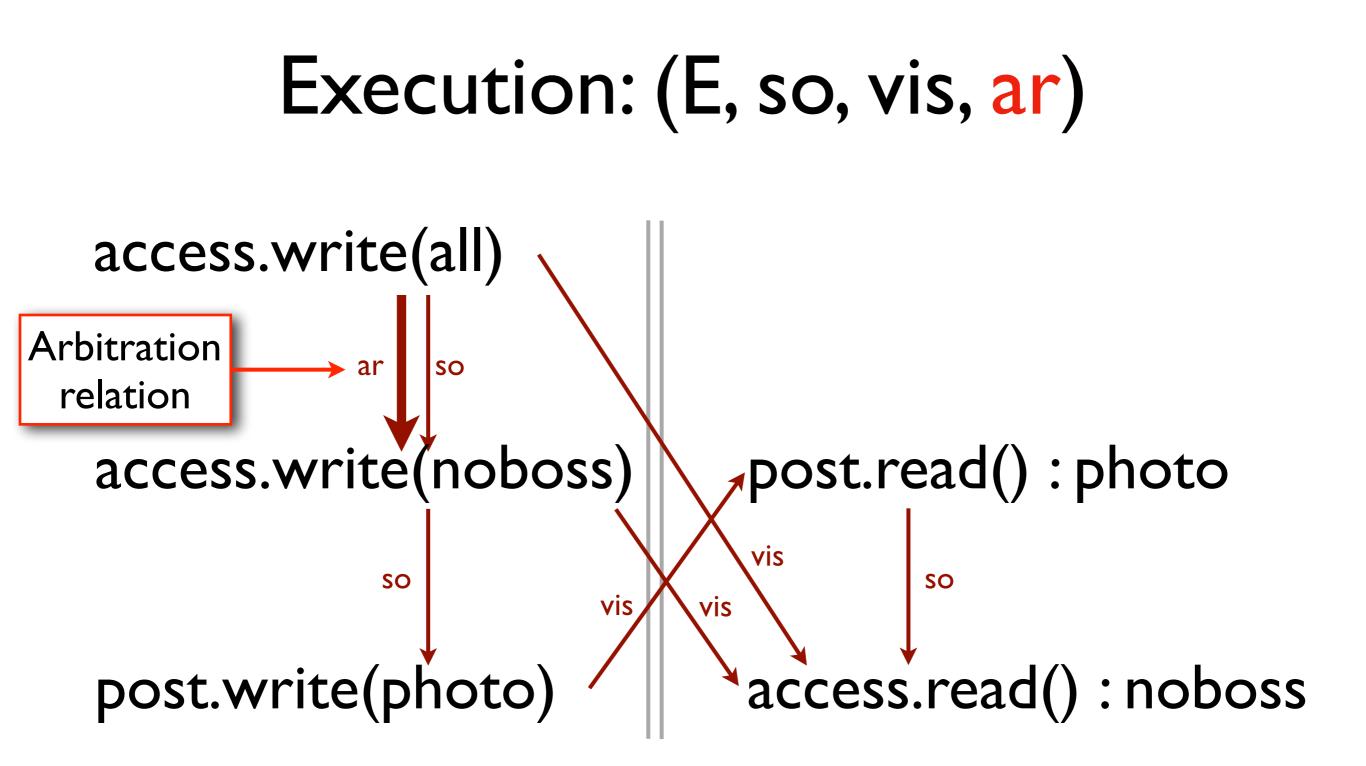
vis is irreflexive and acyclic



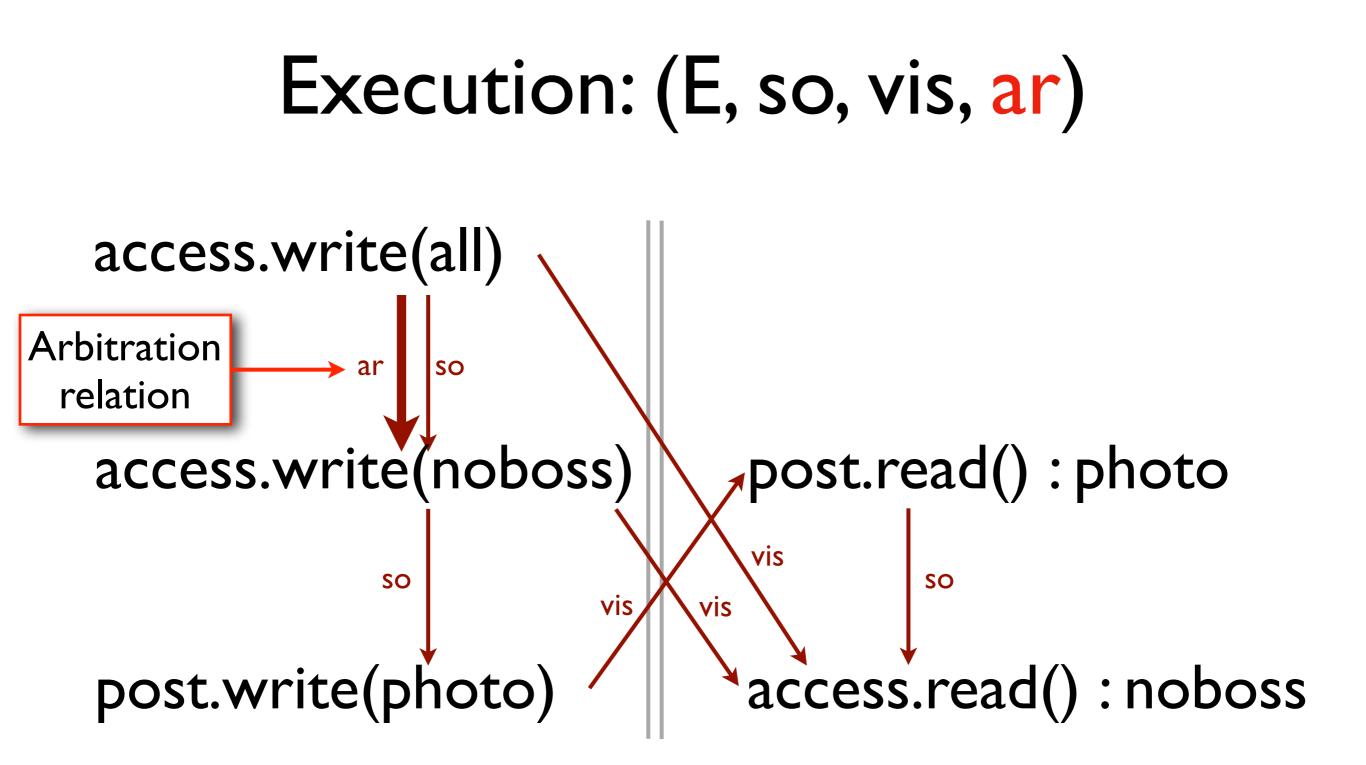
System includes a time-stamping mechanism that can be used in conflict resolution



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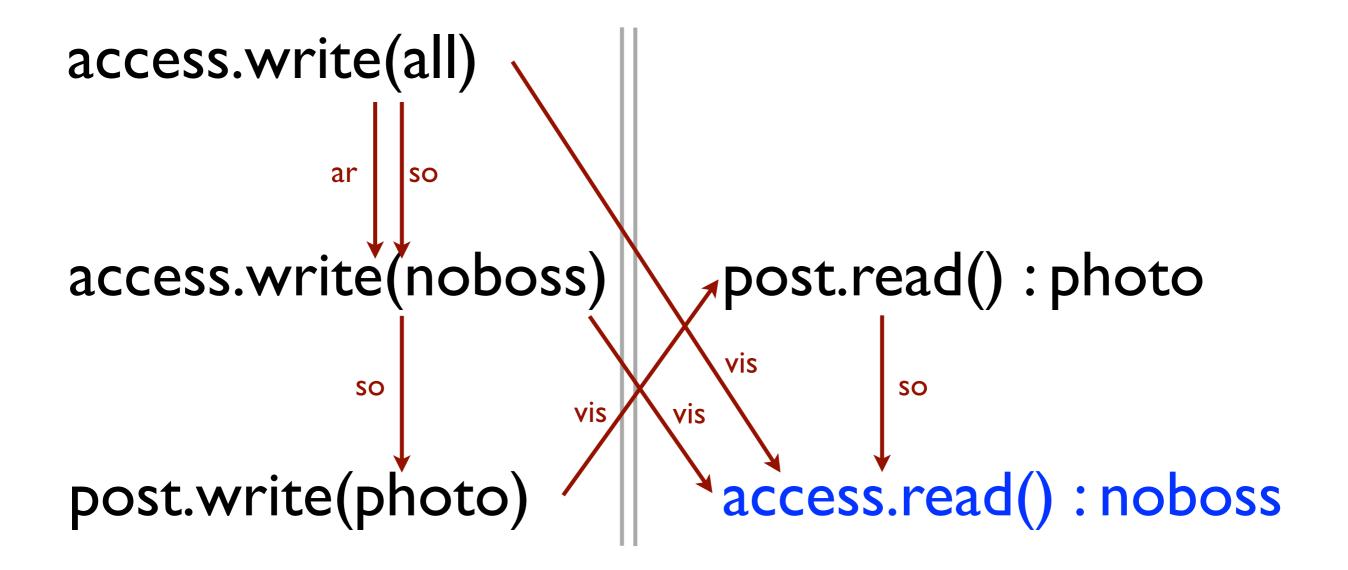


System includes a time-stamping mechanism that can be used in conflict resolution

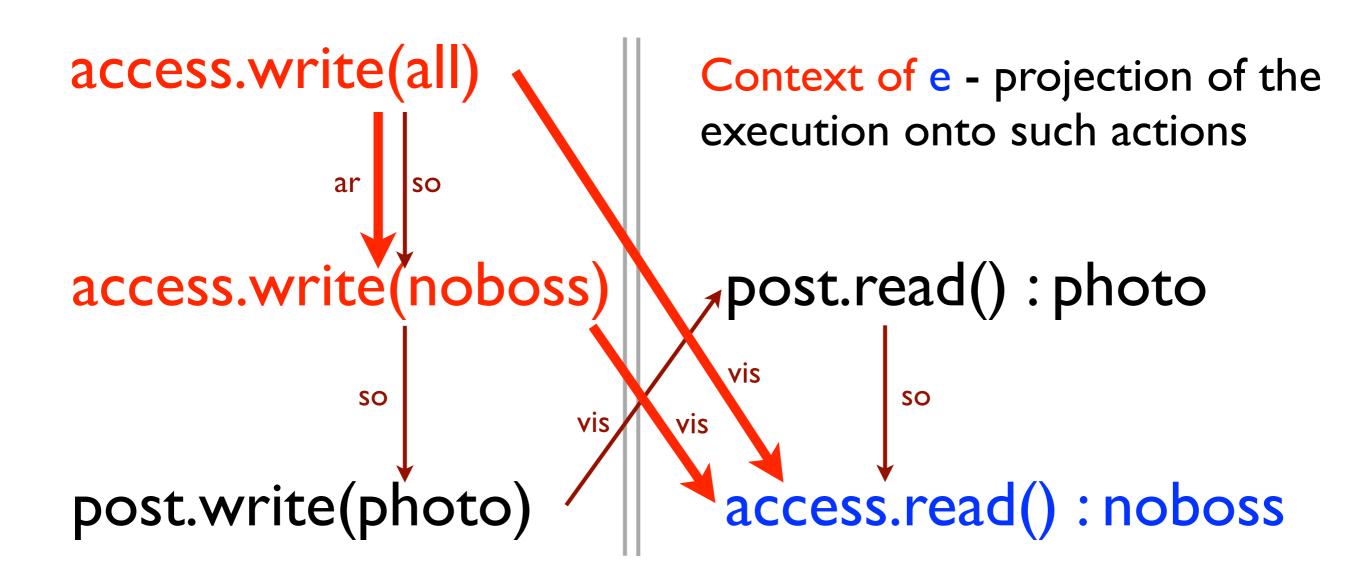


System includes a time-stamping mechanism that can be used in conflict resolution ar is total on E and vis \subseteq ar

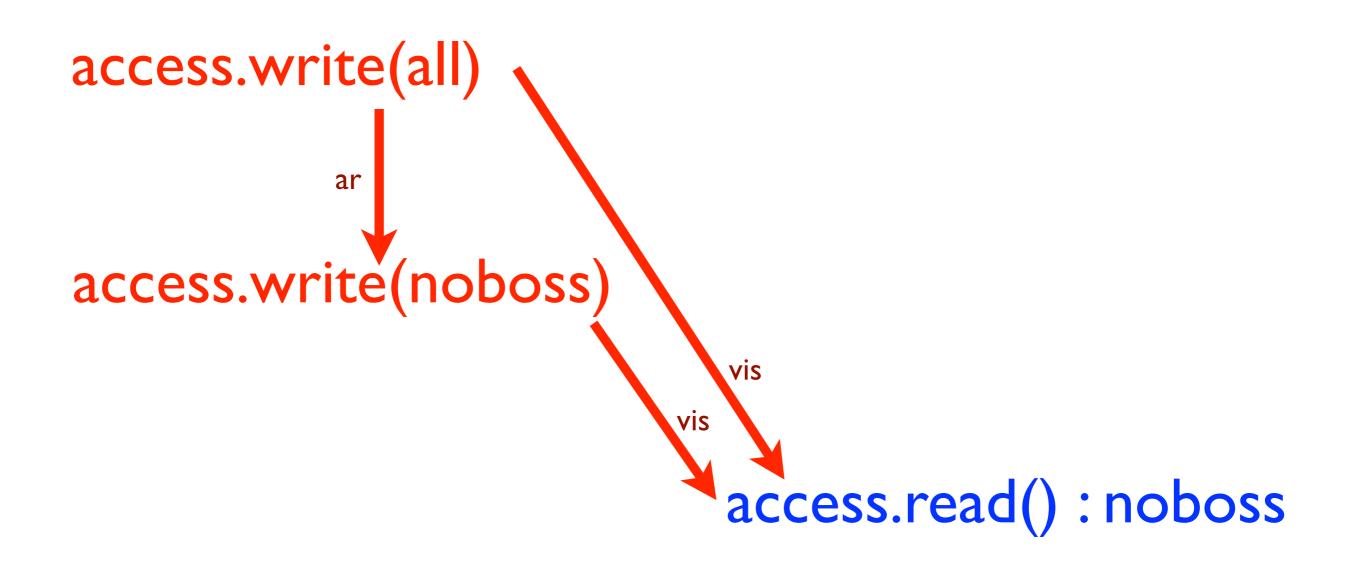
- How do I compute the return value of an event e?
- Only actions on the same object visible to e are important: have been delivered to the replica performing e



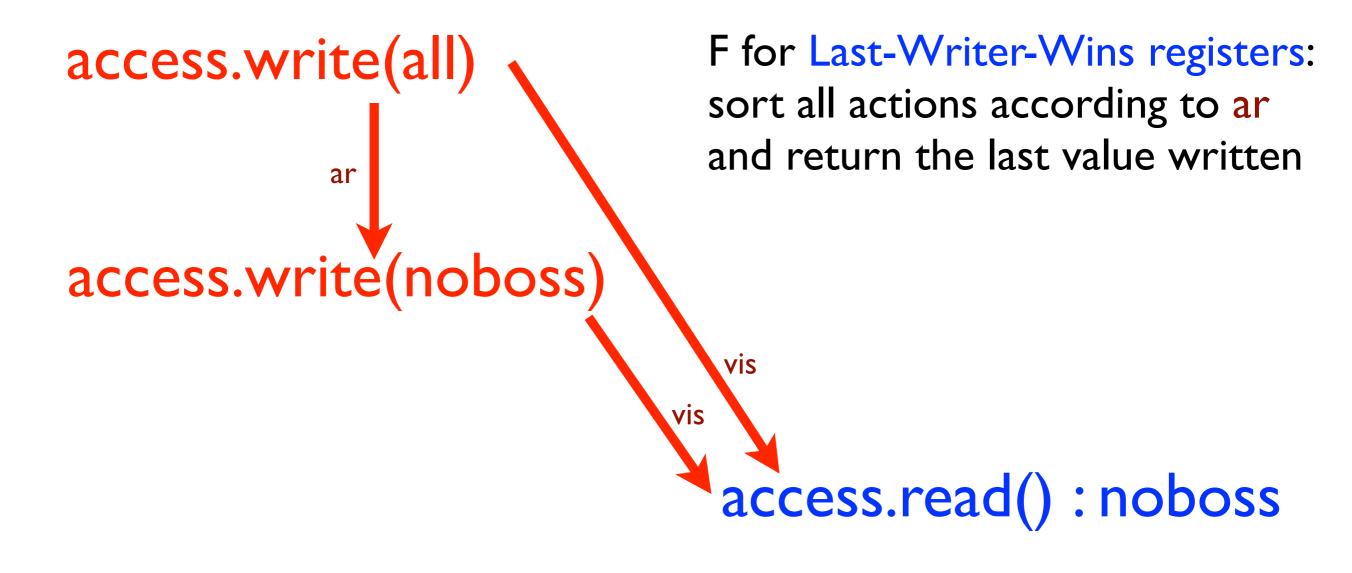
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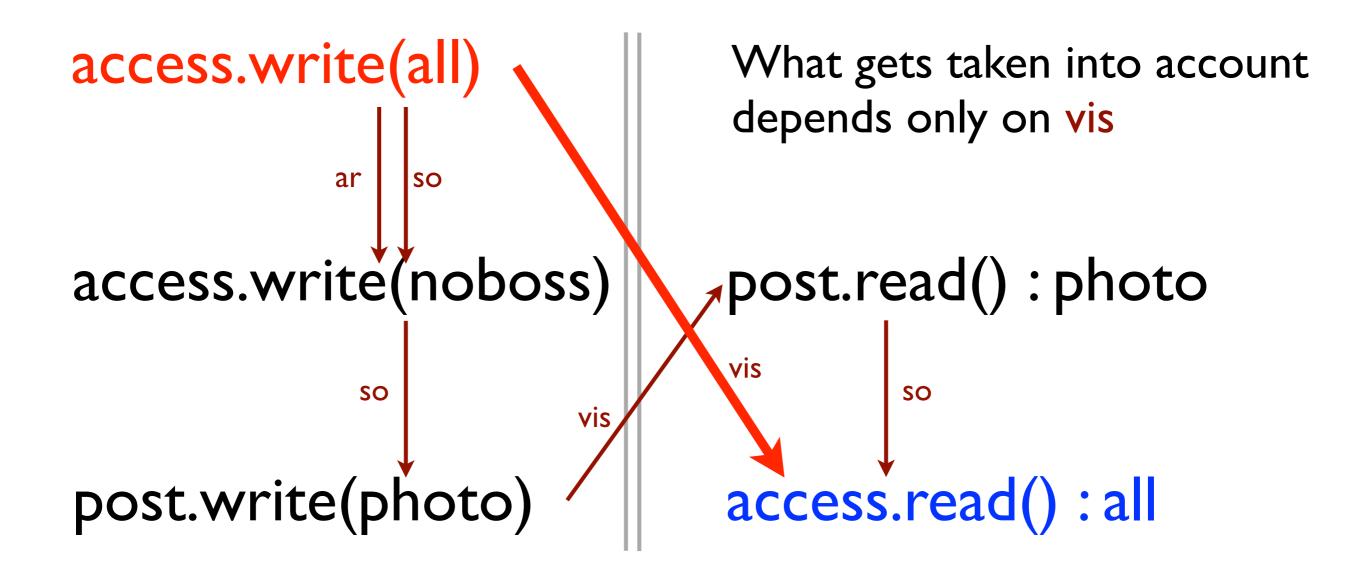
F: context of $e \rightarrow$ return value of e



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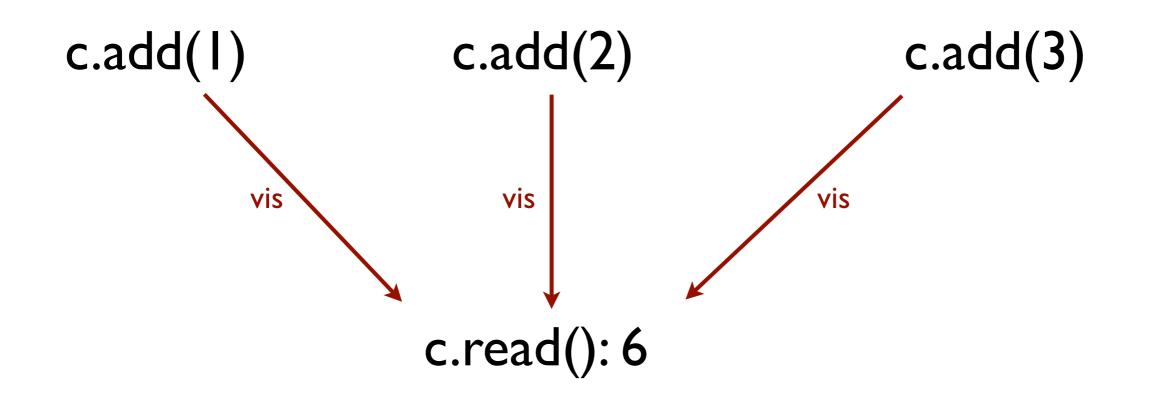


F: context of $e \rightarrow$ return value of e



Counter

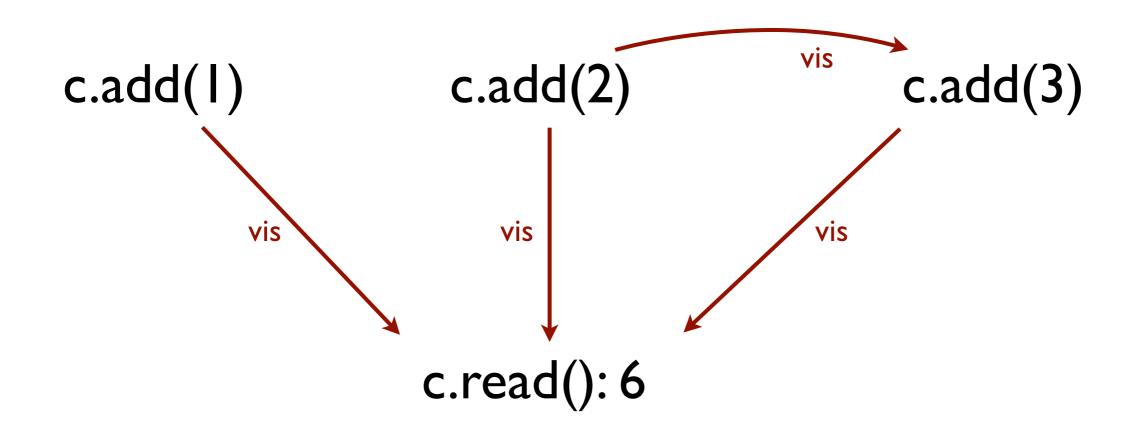
F: context of $e \rightarrow$ return value of e



F: reads return the sum of all additions in the context

Counter

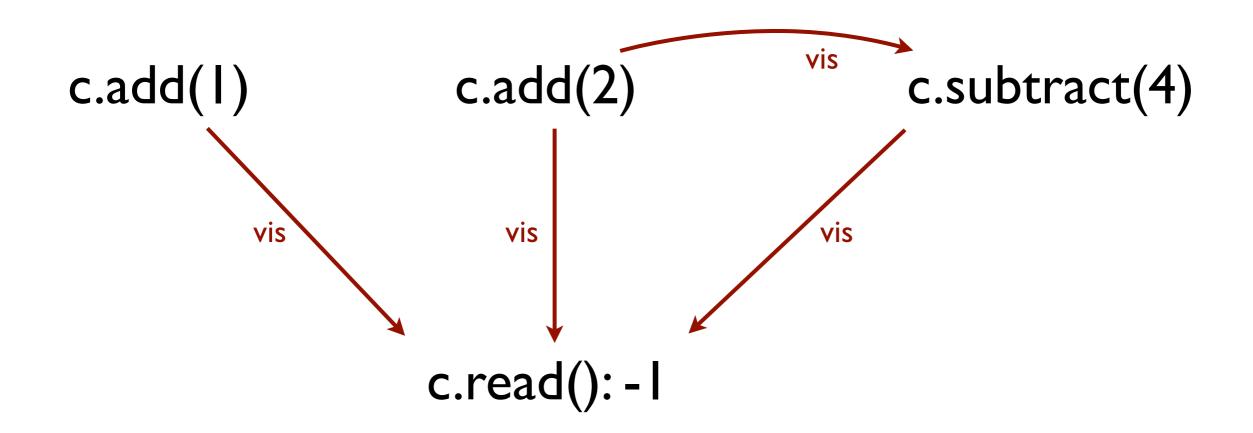
F: context of $e \rightarrow$ return value of e



Relations between events in the context don't matter

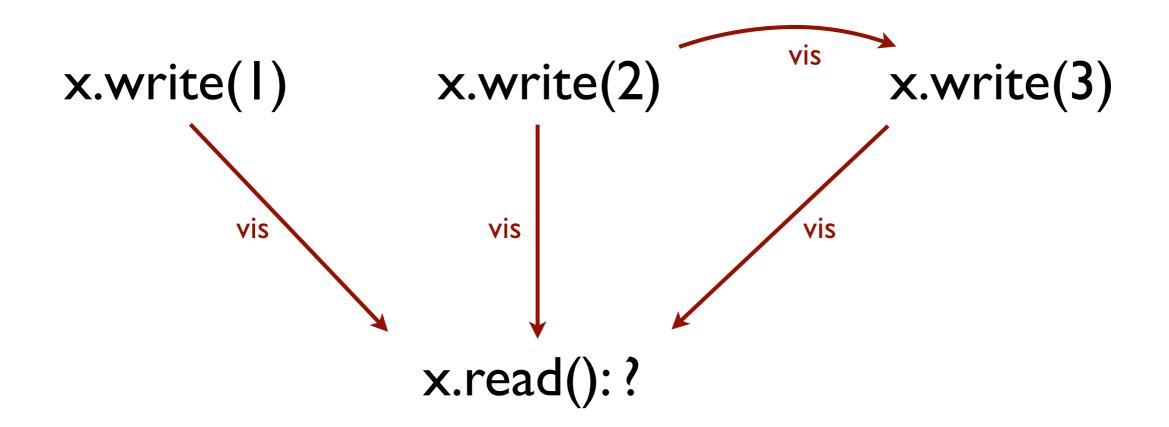
Counter with decrements

F: context of $e \rightarrow$ return value of e

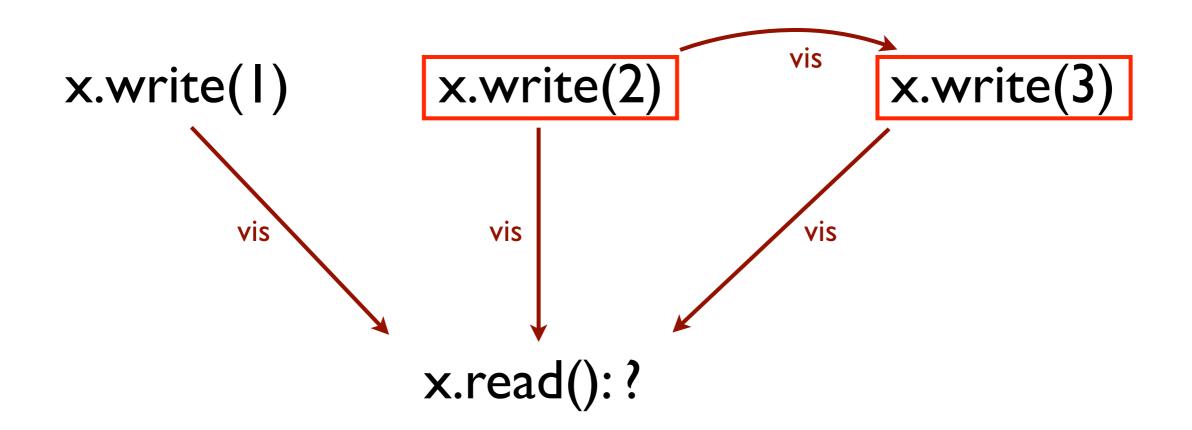


F: reads return additions minus subtractions

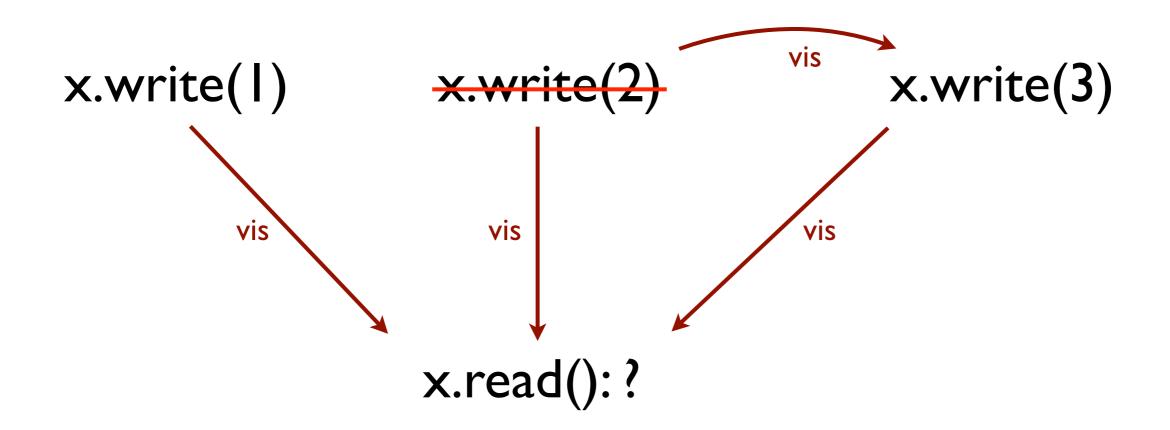
F: context of $e \rightarrow$ return value of e



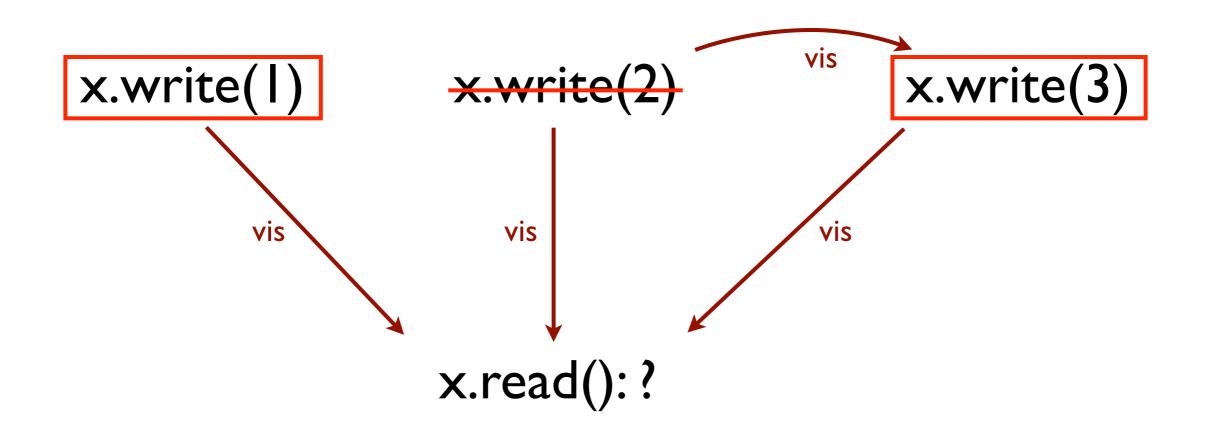
F: context of $e \rightarrow$ return value of e

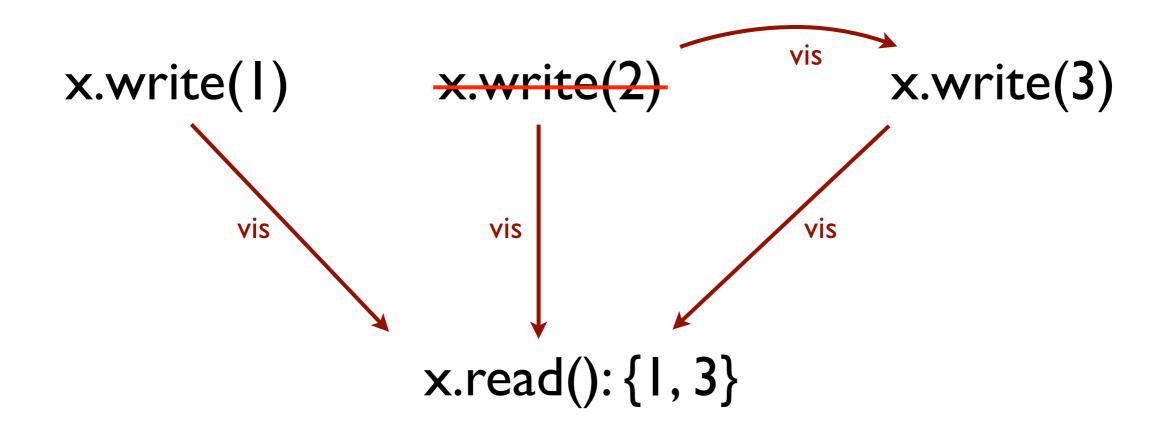


F: context of $e \rightarrow$ return value of e

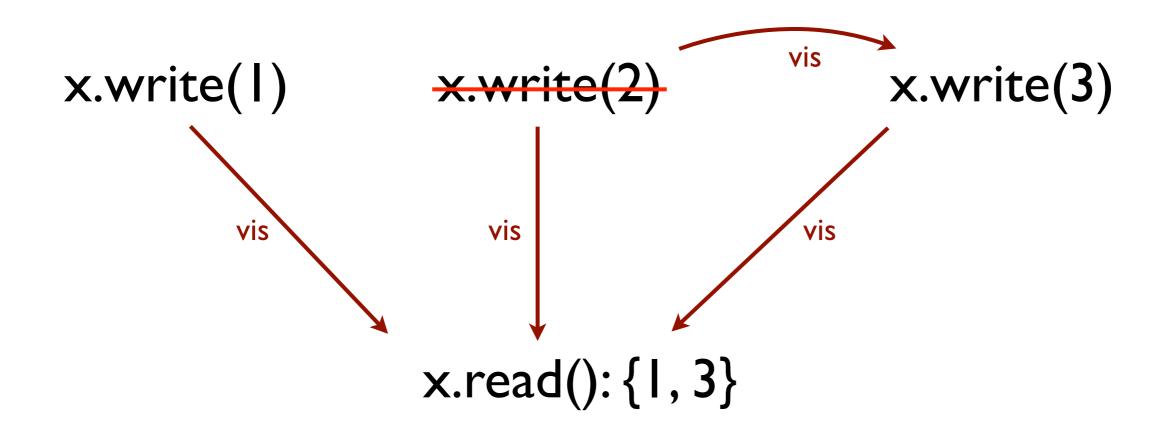


F: context of $e \rightarrow$ return value of e



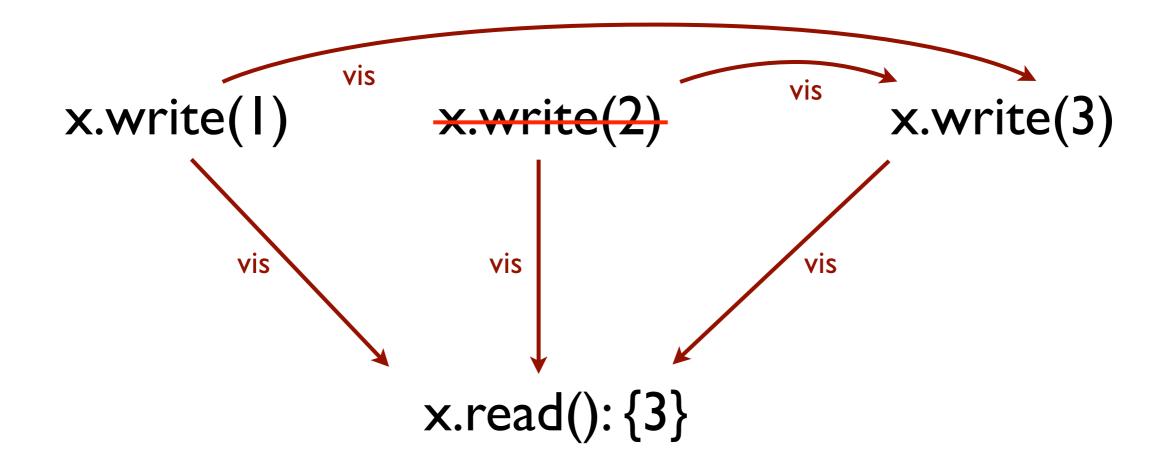


F: context of $e \rightarrow$ return value of e

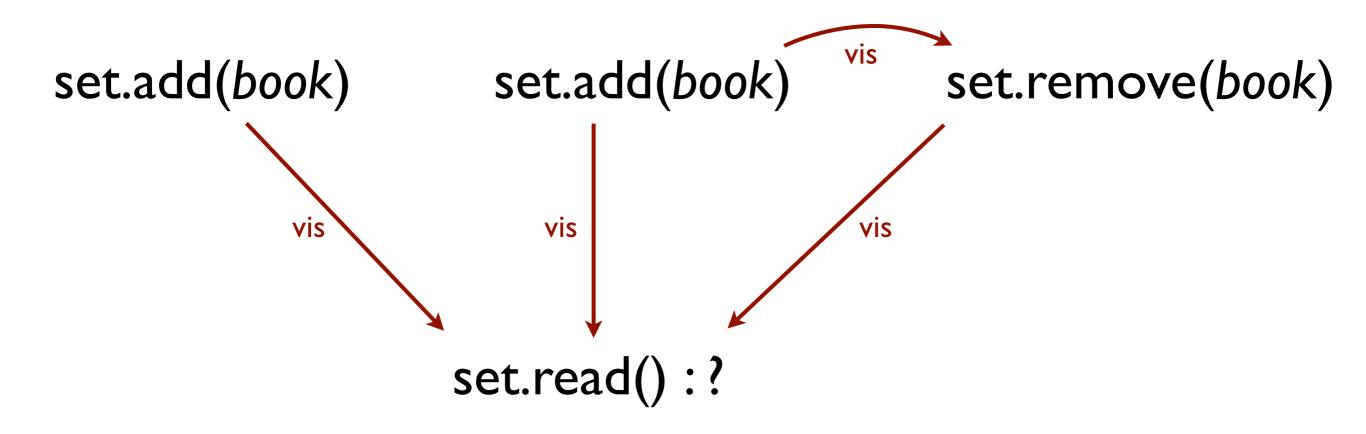


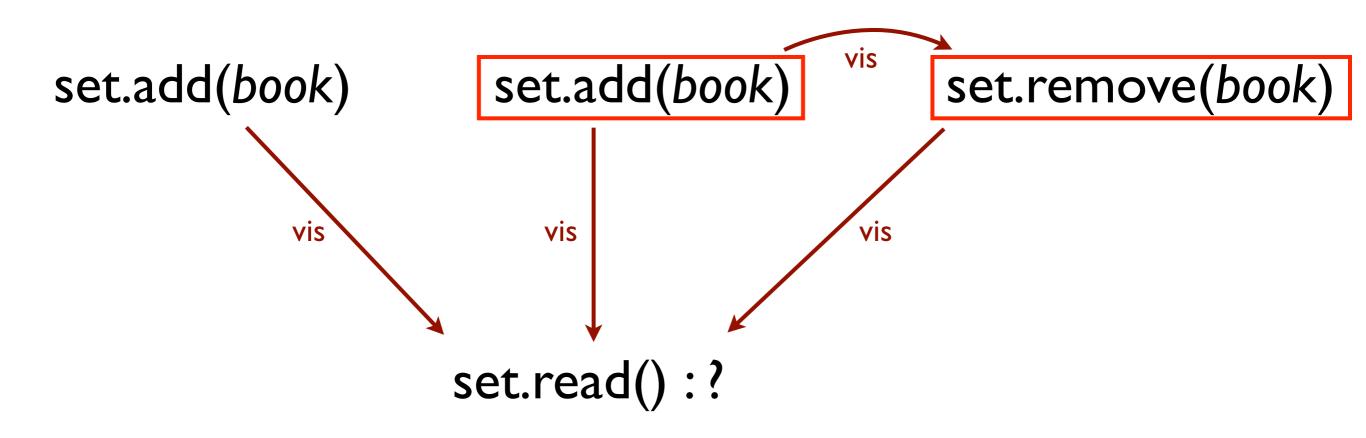
F: discard all writes seen by a write

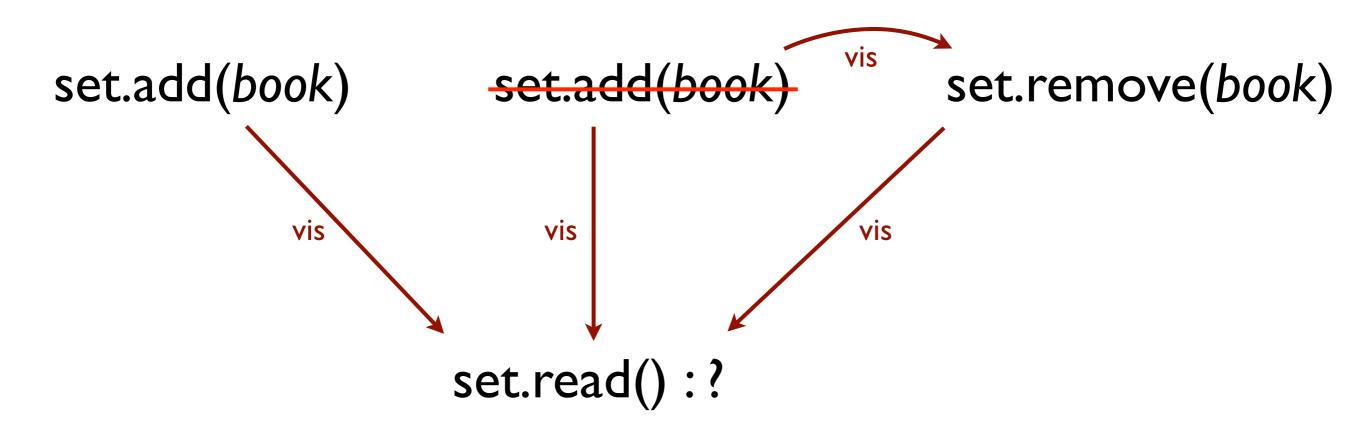
F: context of $e \rightarrow$ return value of e

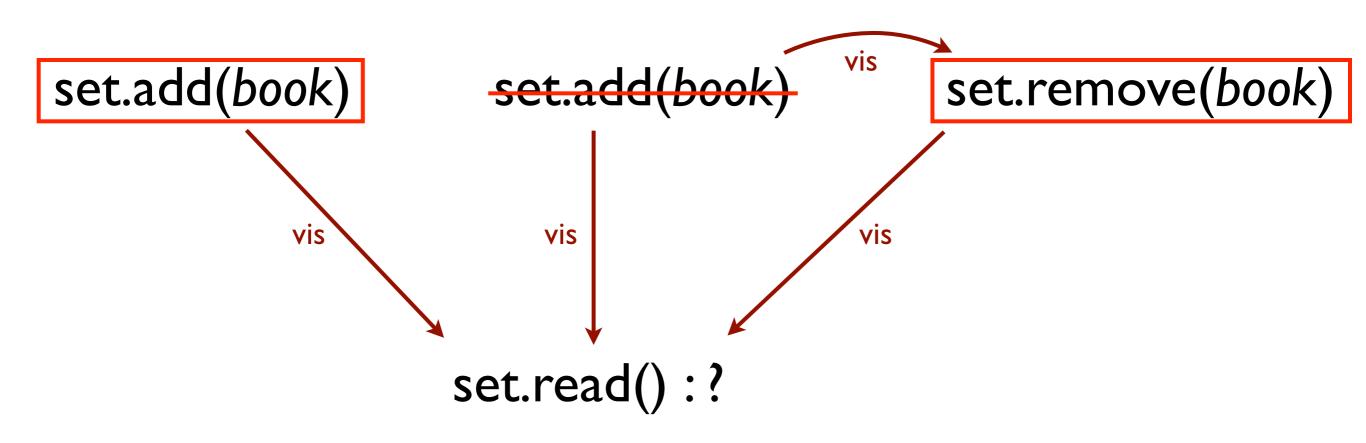


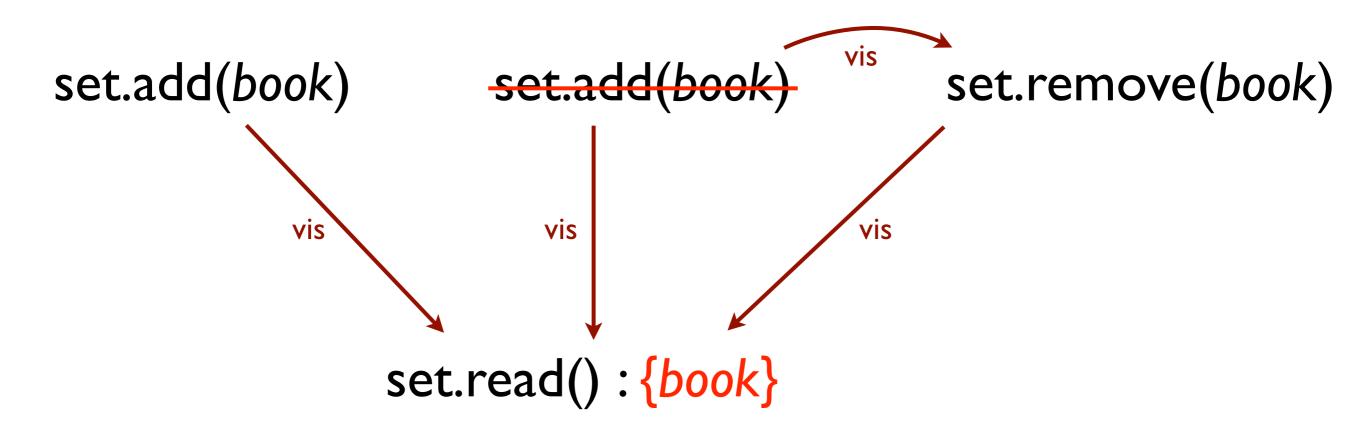
F: discard all writes seen by a write



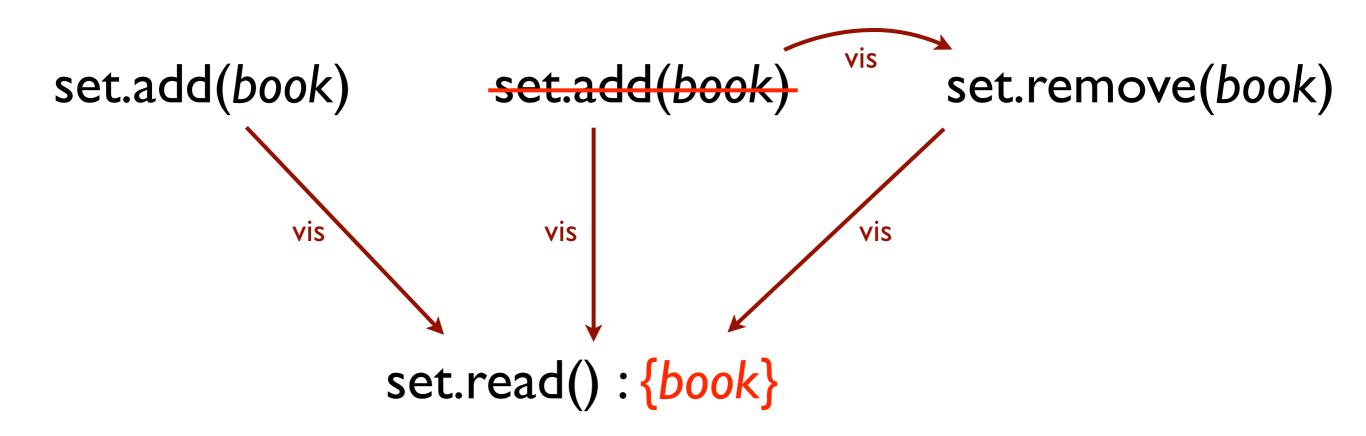






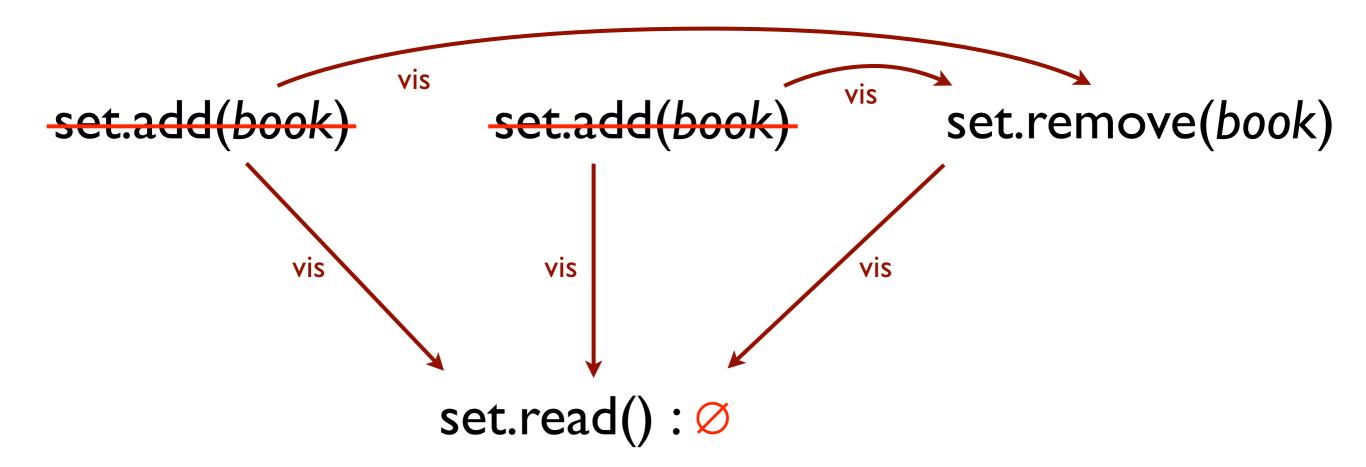


F: context of $e \rightarrow$ return value of e



F: cancel all adds seen by a remove

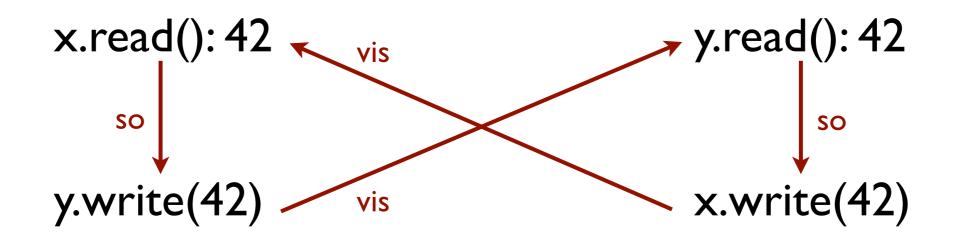
F: context of $e \rightarrow$ return value of e



F: cancel all adds seen by a remove

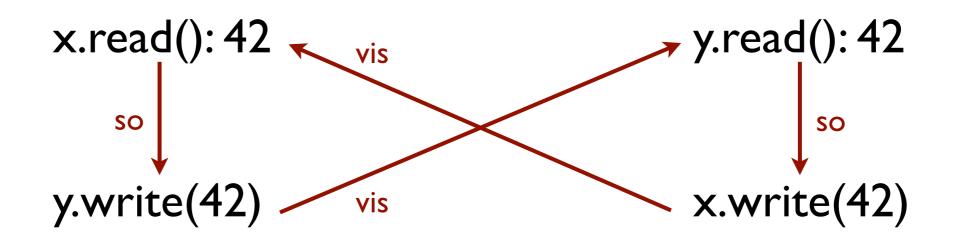
F: context of $e \rightarrow$ return value of e

"No causal cycles" axiom



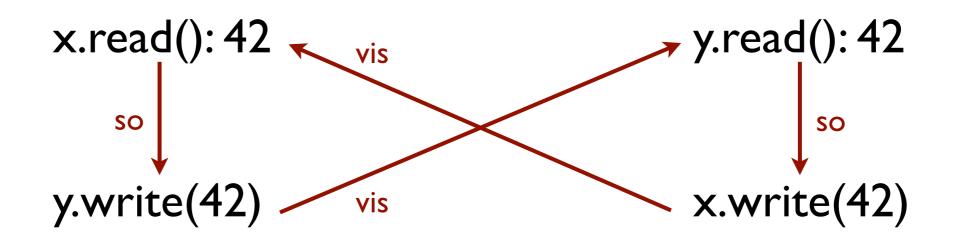
- so \cup vis is acyclic: no causal cycles/out-of-thin-air values
- so and vis consistent with execution order

"No causal cycles" axiom

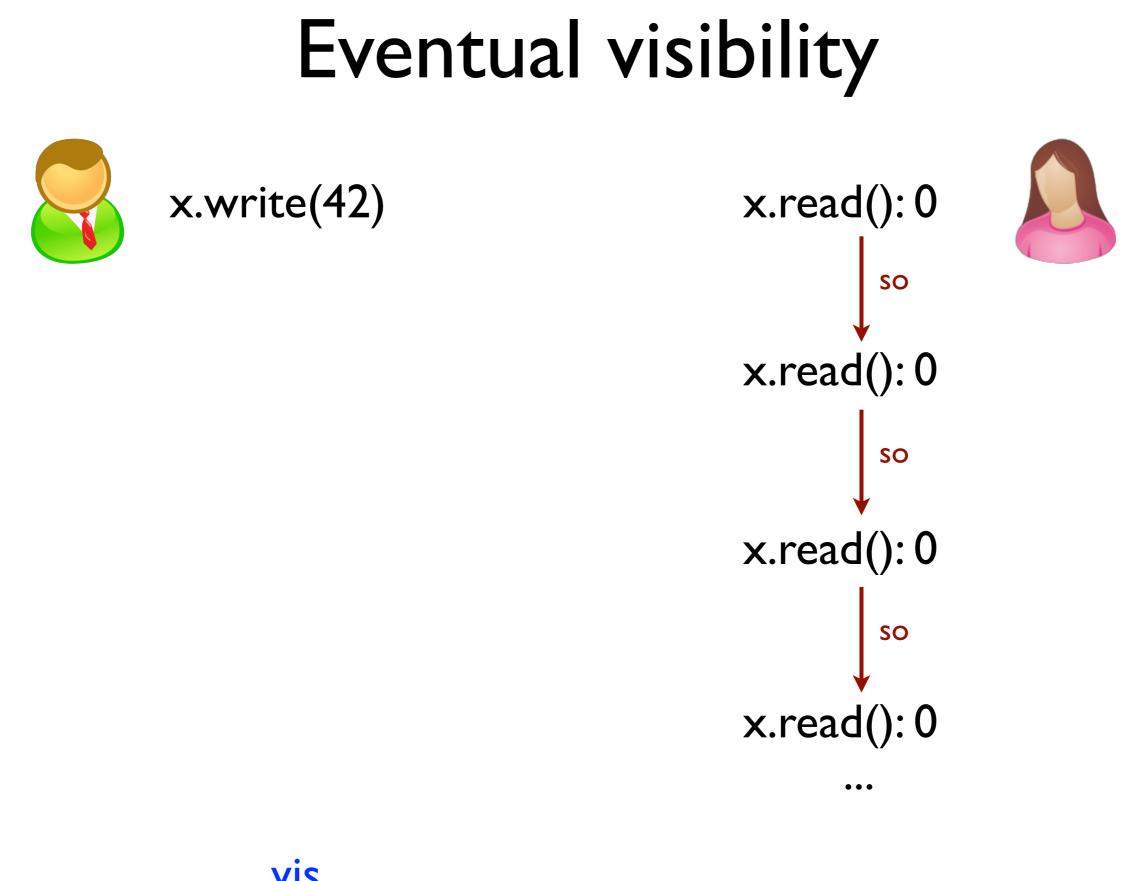


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- Could result from speculative execution, uncommon in distributed systems

"No causal cycles" axiom

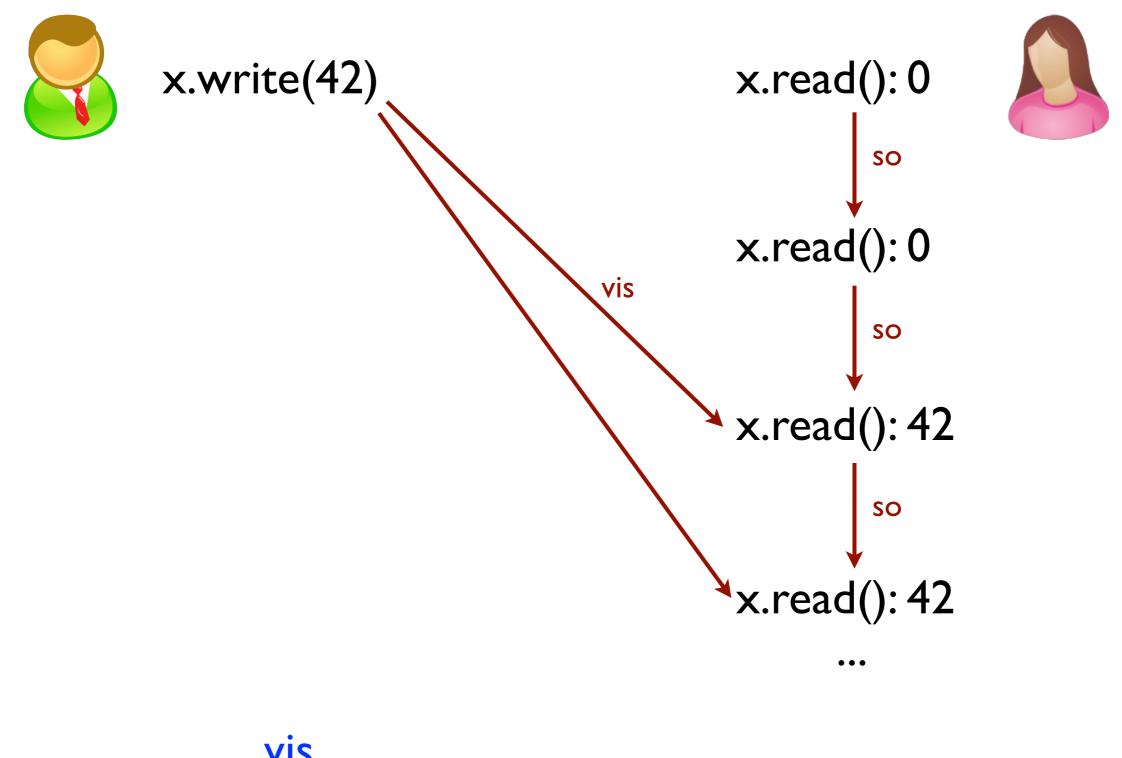


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- so and vis consistent with execution order
- Could result from speculative execution, uncommon in distributed systems
- Some forms allowed by shared-memory models (ARM, C++, Java): defining semantics is an open problem



 $\forall e \in E. e \xrightarrow{vis} f$ for all but finitely many $f \in E$

Eventual visibility



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Eventual consistency summary

The set of histories (E, so) such that for some vis, ar:

- Return values consistent with data type specs: $\forall e \in E. rval(e) = F_{type(obj(e))}(context(e))$
- No causal cycles: so \cup vis is acyclic
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Quiescent consistency: if no new updates are made to the database, then replicas will eventually converge to the same state

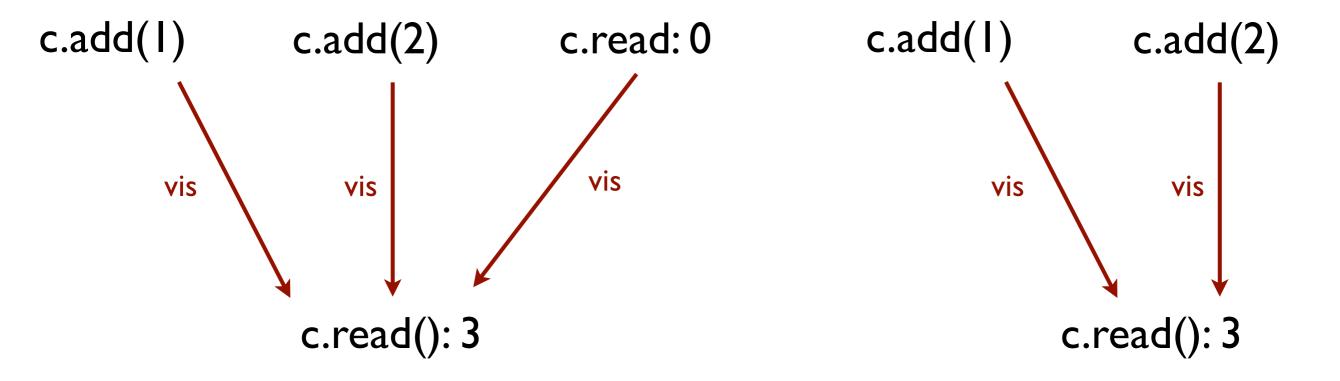
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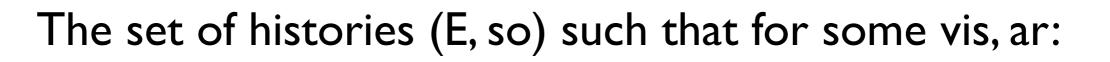
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Stronger than quiescent consistency, but still weak

Strengthen consistency by adding additional axioms on vis and ar



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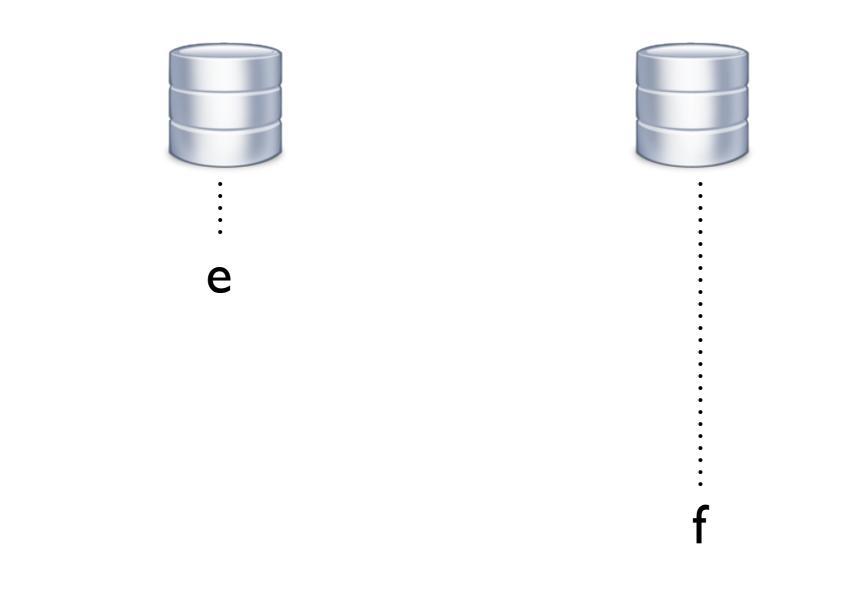
- Proofs depend on replicated data types
- Example: replicated counters and last-writer-wins registers
- There are also generic proof techniques that work for whole classes of data types

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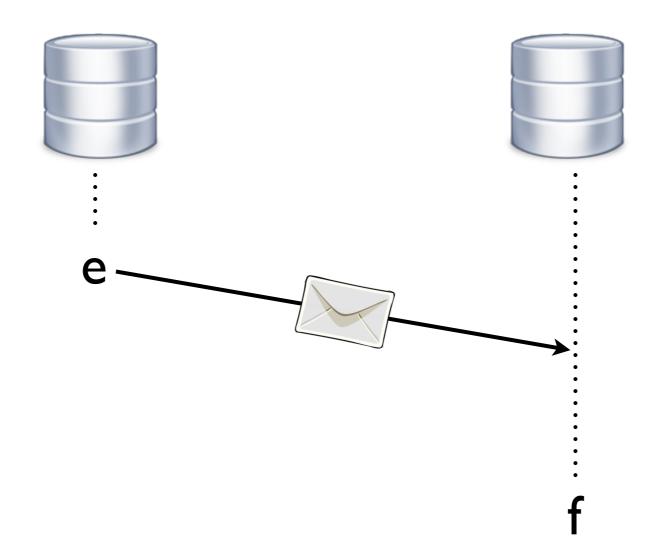
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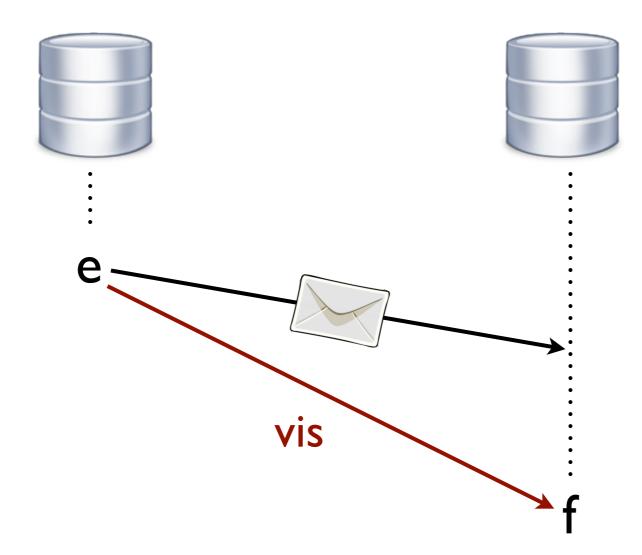
Constructing vis



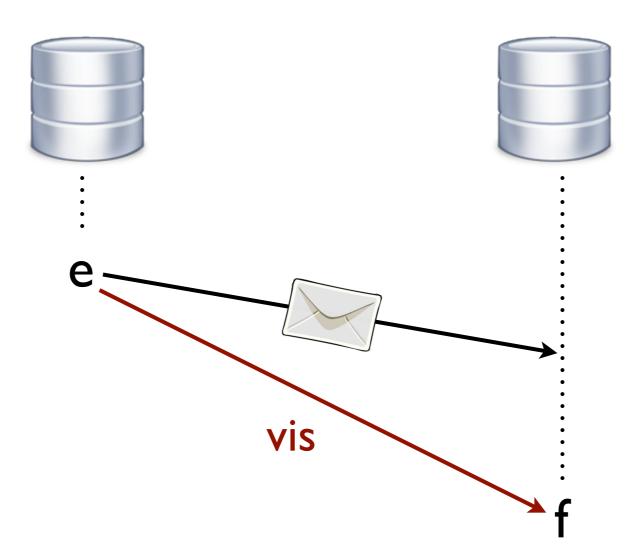
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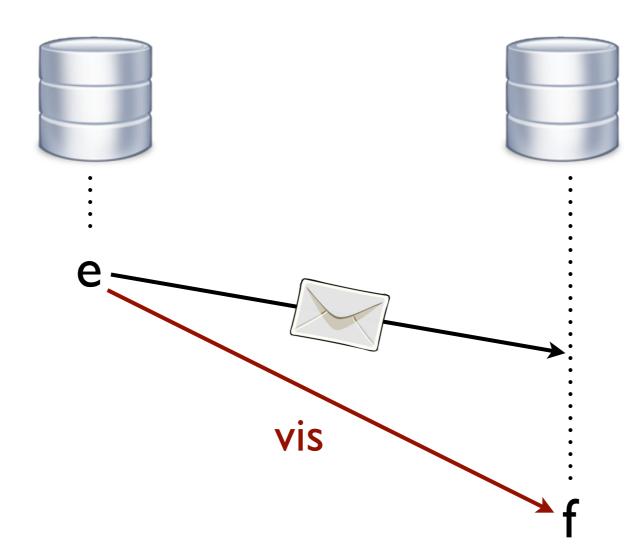
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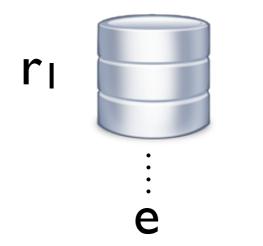
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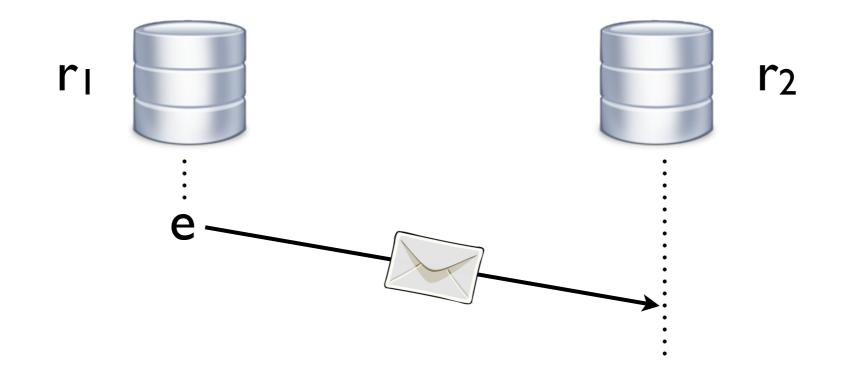
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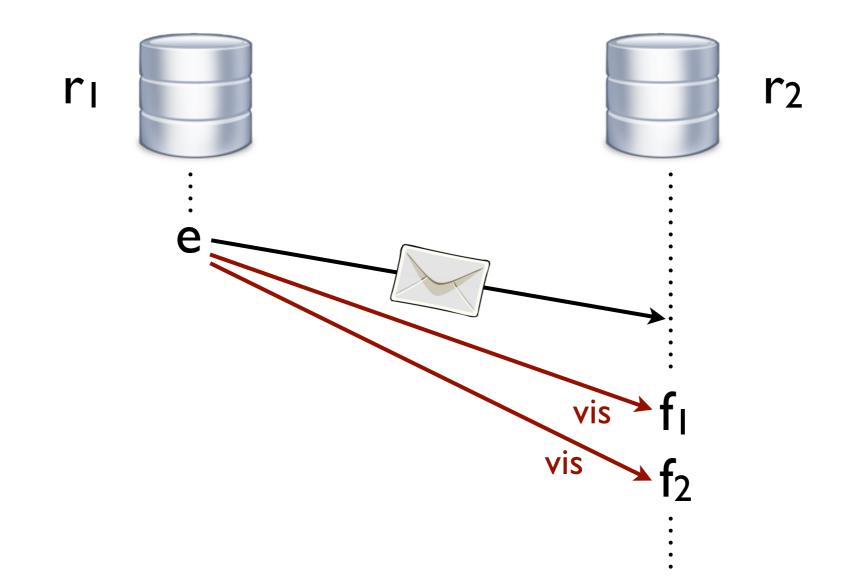
 $e \xrightarrow{vis} f \lor e \xrightarrow{so} f \Longrightarrow e$ was issued before f in the operational execution



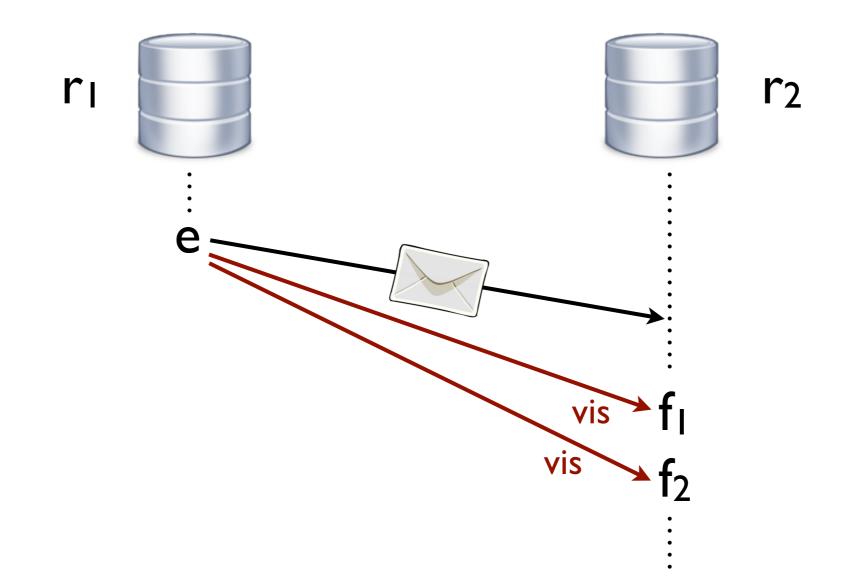




• Channels are reliable (every partition eventually heals) \implies the effector of e is eventually delivered to r_2

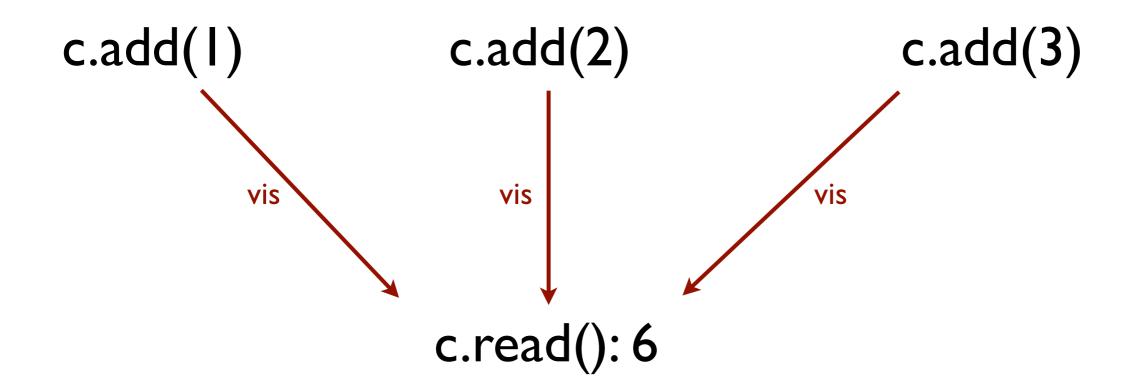


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- True for any replica \implies only finitely many events don't see e

 $\forall e \in E. rval(e) = F_{type(obj(e))}(context(e))$

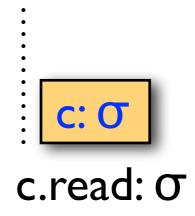


F: reads return the sum of all additions in the context

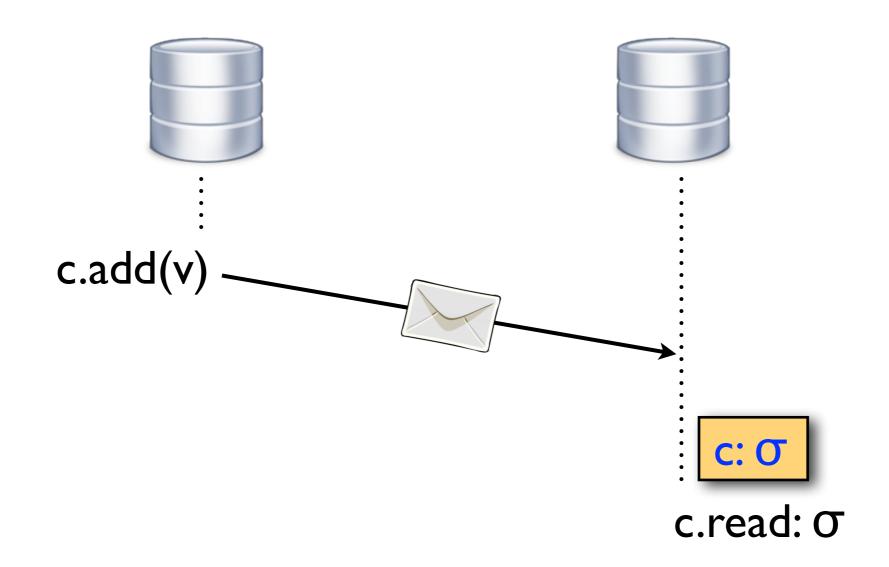




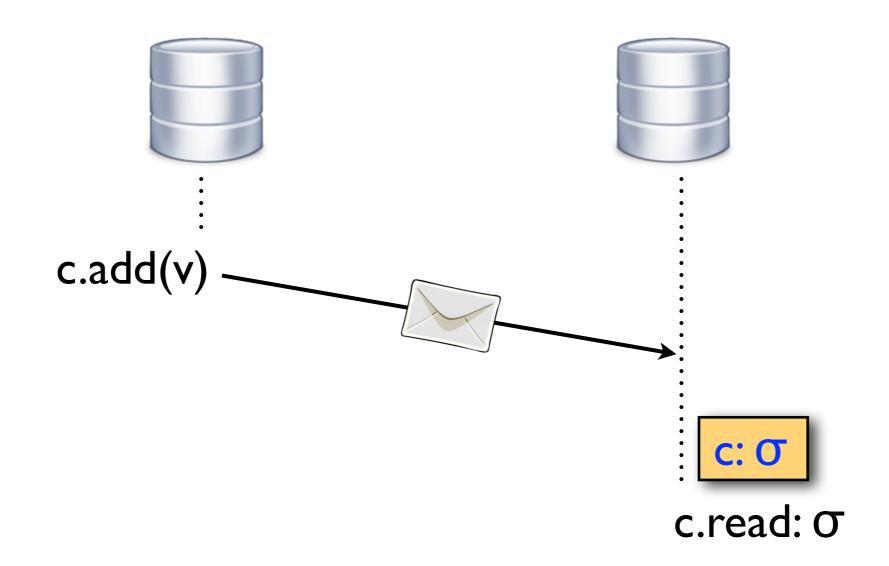




A read returns the value of the counter at the replica: $[read()]_{val}(\sigma) = \sigma$

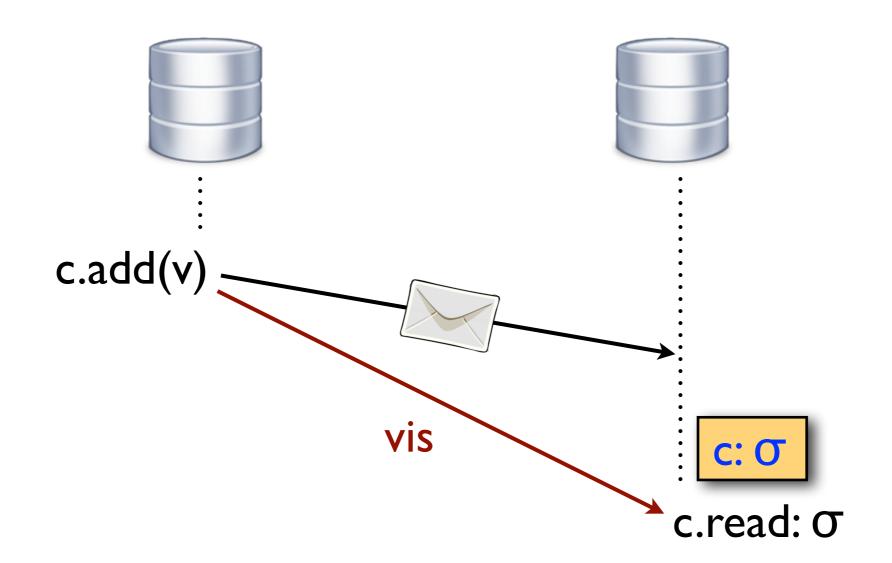


Invariant: the value of a counter at a replica is the sum of all increments of the counter delivered to it



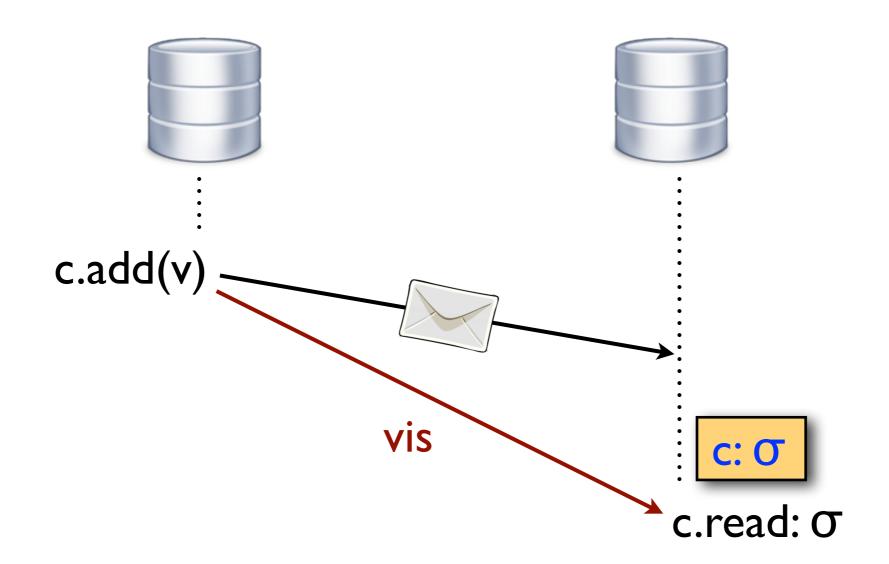
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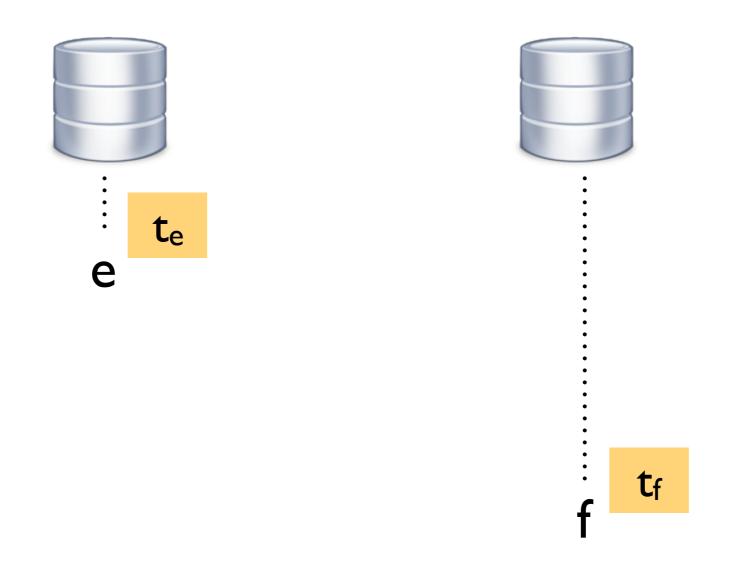
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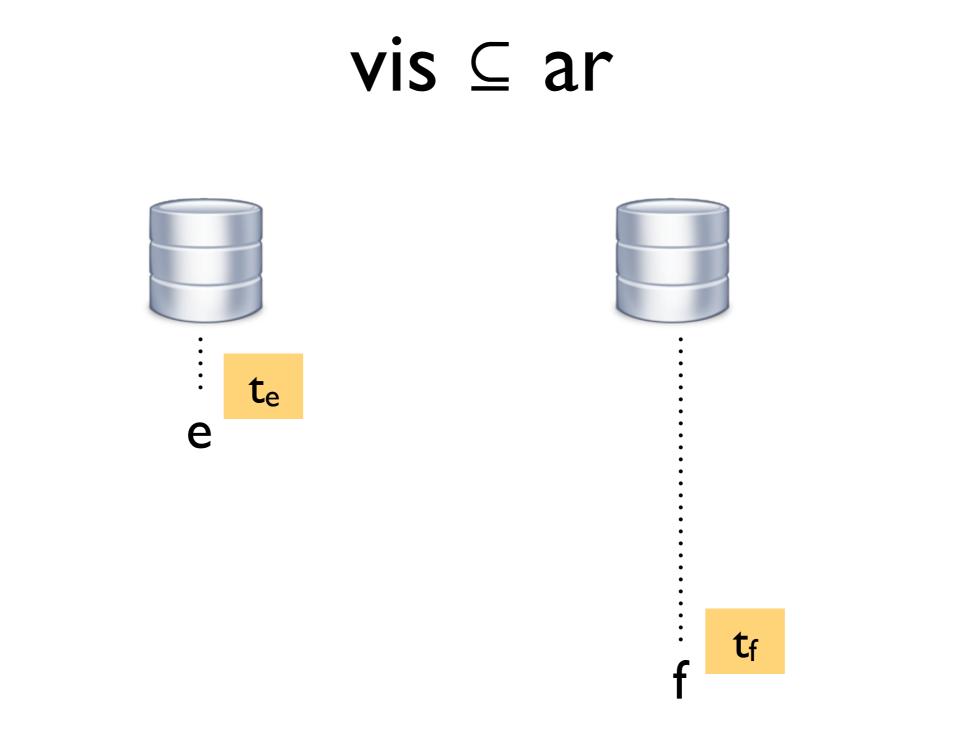
= increments visible to the read, QED.

Constructing ar



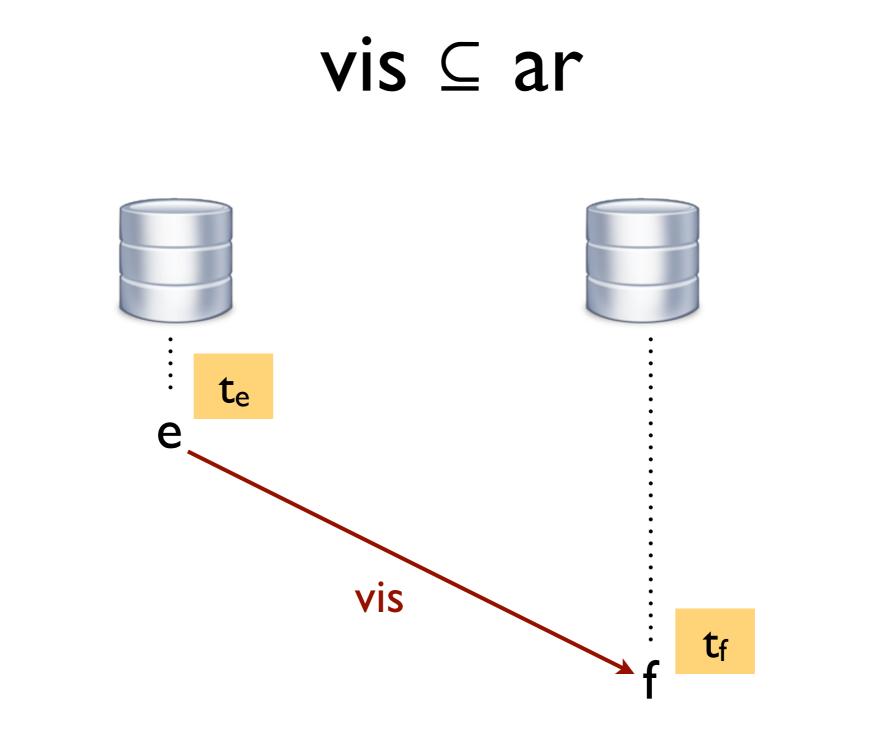
Every event e gets assigned a timestamp $t_{\rm e}$ from a logical Lamport clock

 $e \stackrel{ar}{\longrightarrow} f \Longleftrightarrow t_e \leq t_f$

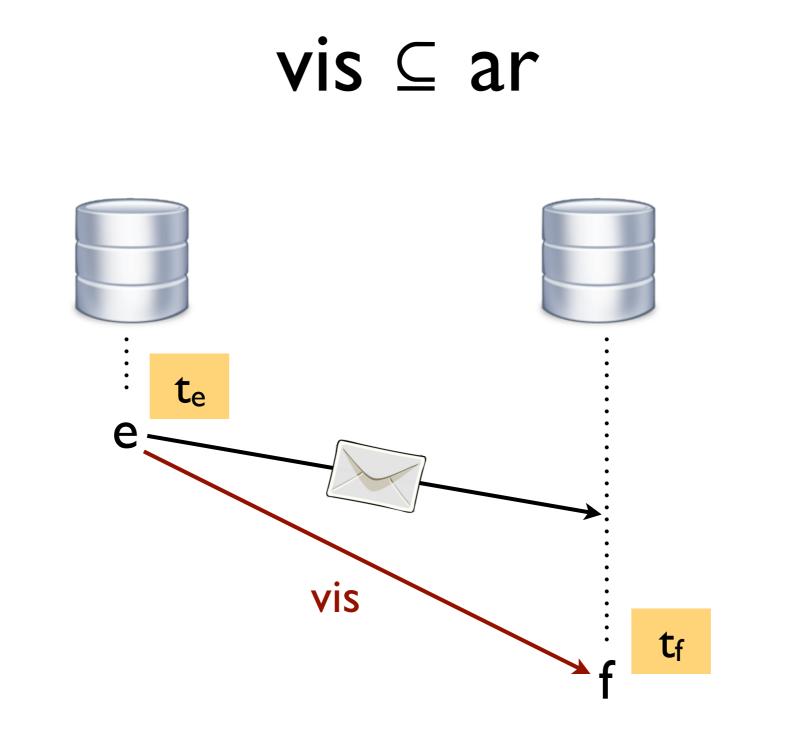


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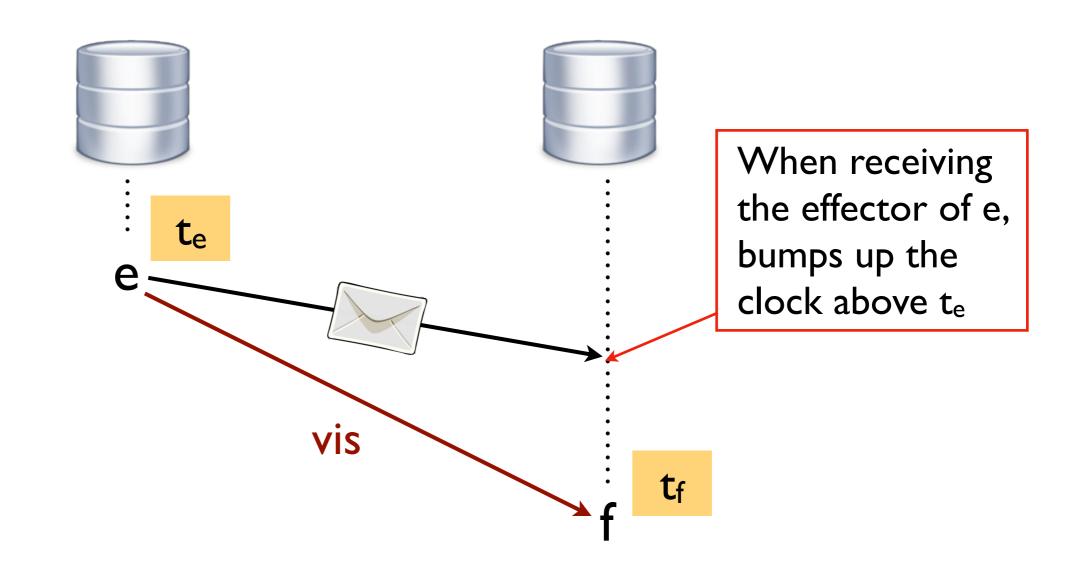


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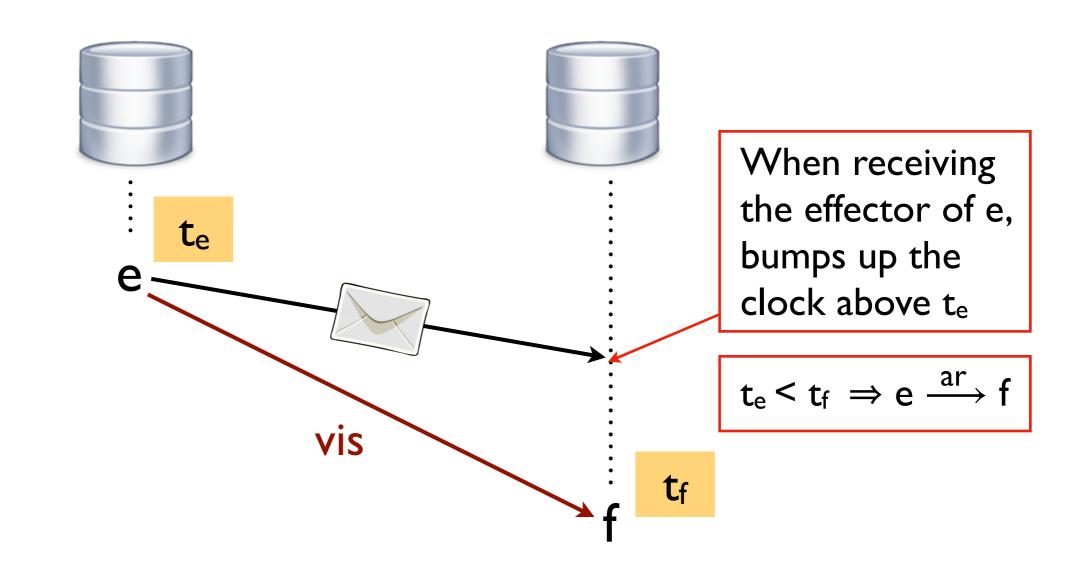
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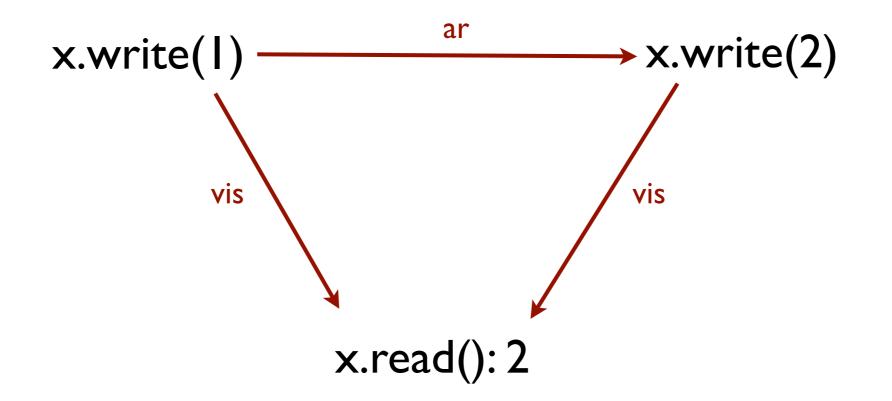
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Every event e gets assigned a timestamp $t_{\rm e}$ from a logical Lamport clock

Correctness of registers

 $\forall e \in E. rval(e) = F_{type(obj(e))}(context(e))$



F: reads return the last value in ar

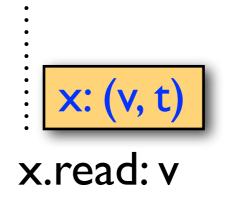
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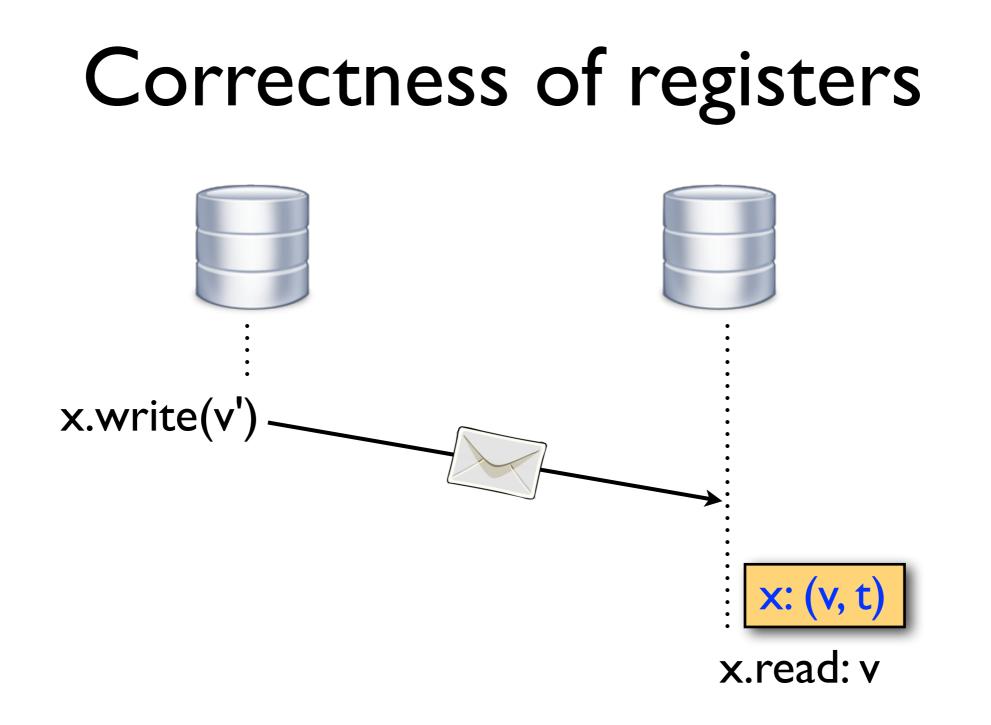
x.read: ?

Correctness of registers

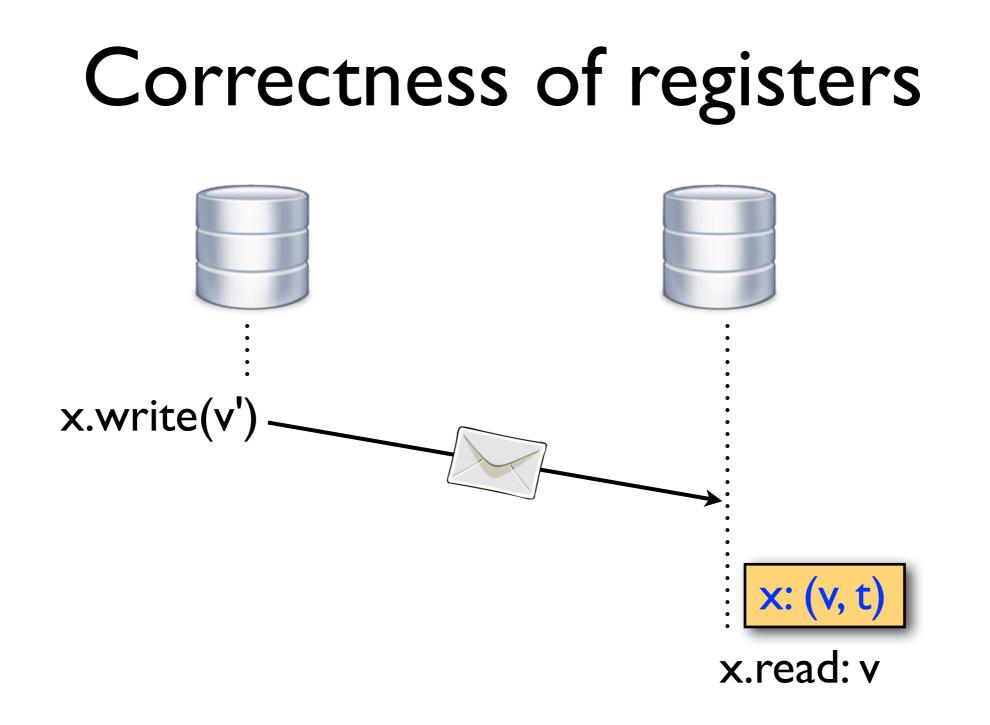




A read returns the value part of the register at the replica: $[read()]_{val}(v, t) = v$

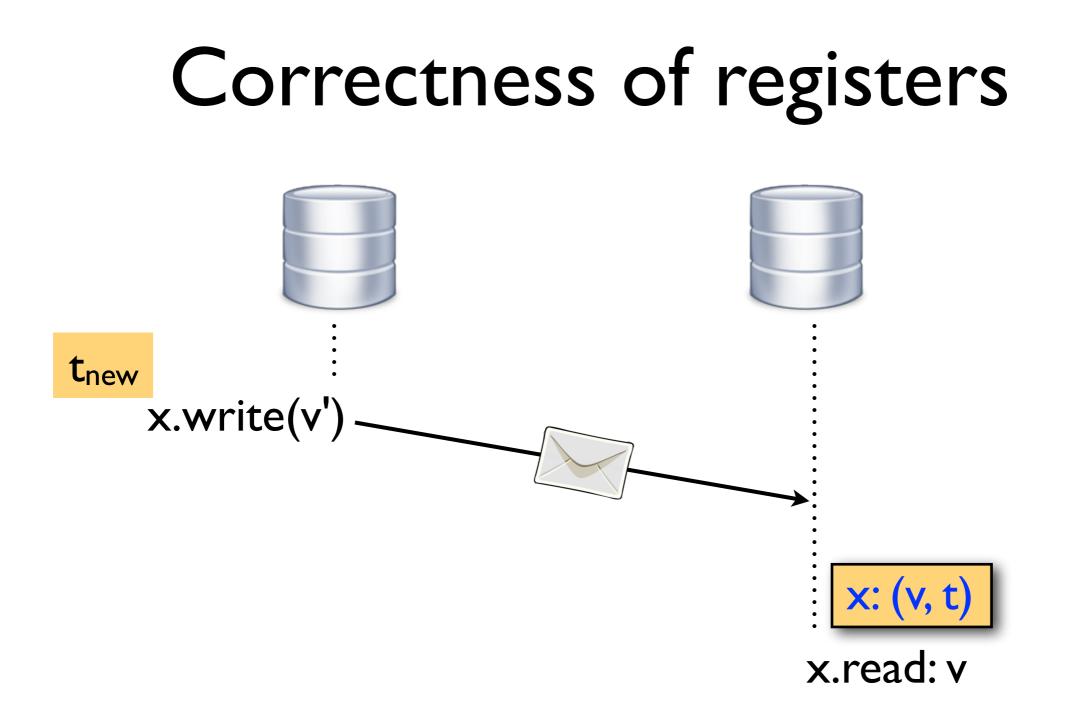


Invariant: the value of a register at a replica is the one with the highest timestamp out of all delivered writes



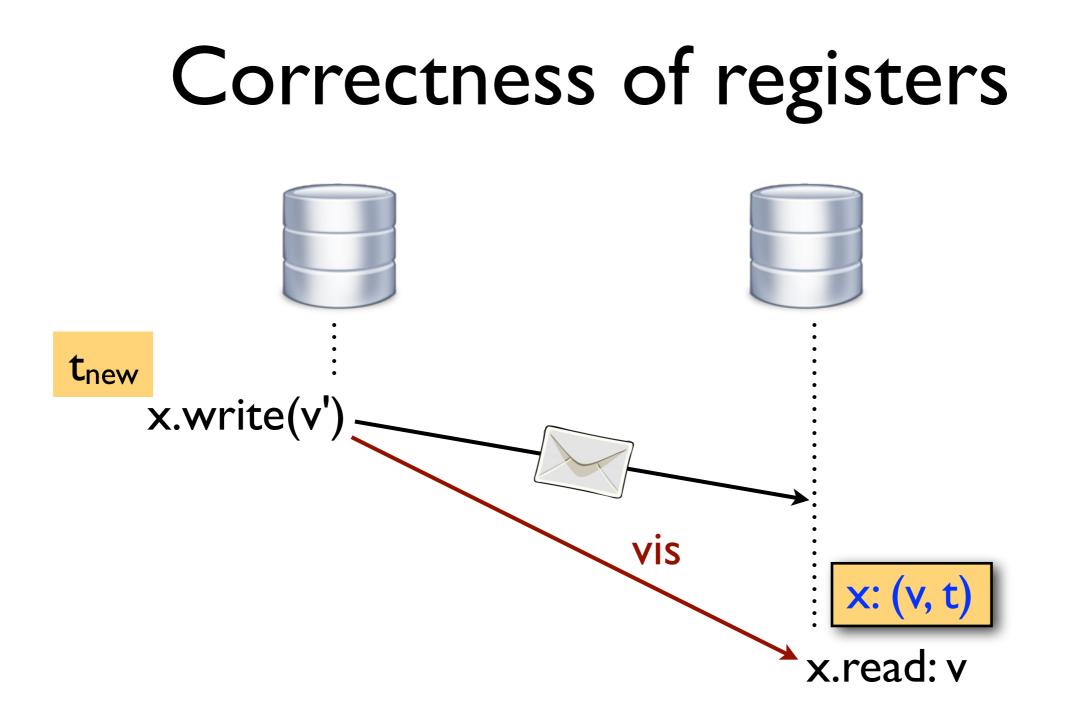
Invariant: the value of a register at a replica is the one with the highest timestamp out of all delivered writes

 $[vrite(v_{new})]_{eff}(v, t) = let (t_{new} = newUniqueTS()) in$ $\lambda(v', t'). if t_{new} > t' then (v_{new}, t_{new}) else (v, t)$



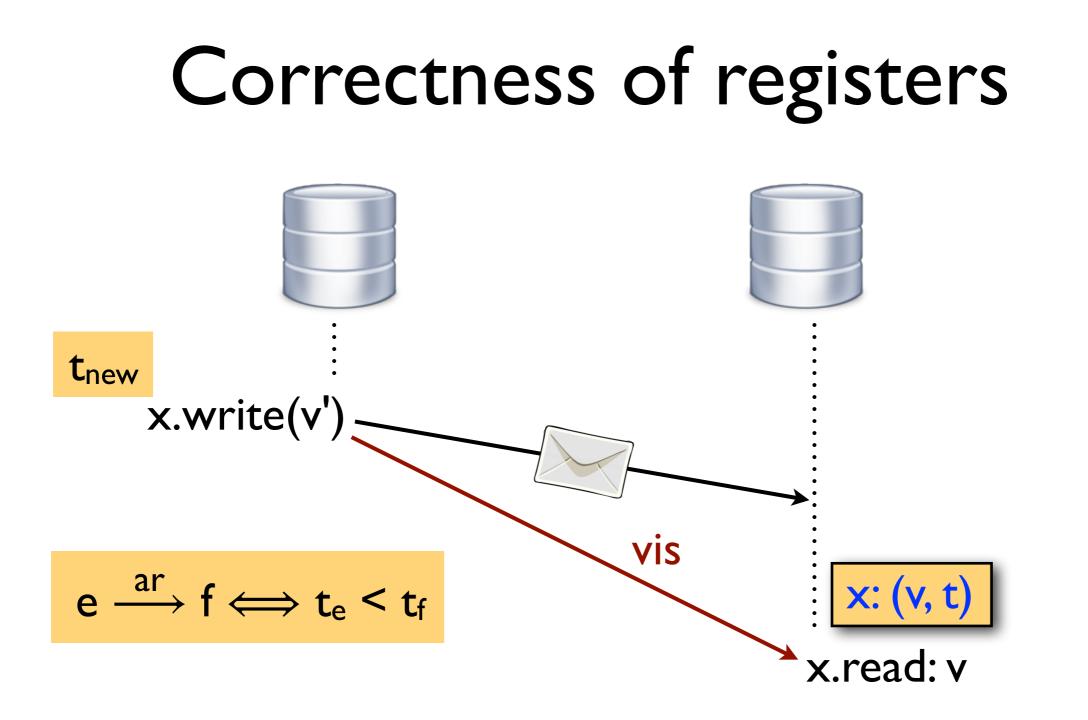
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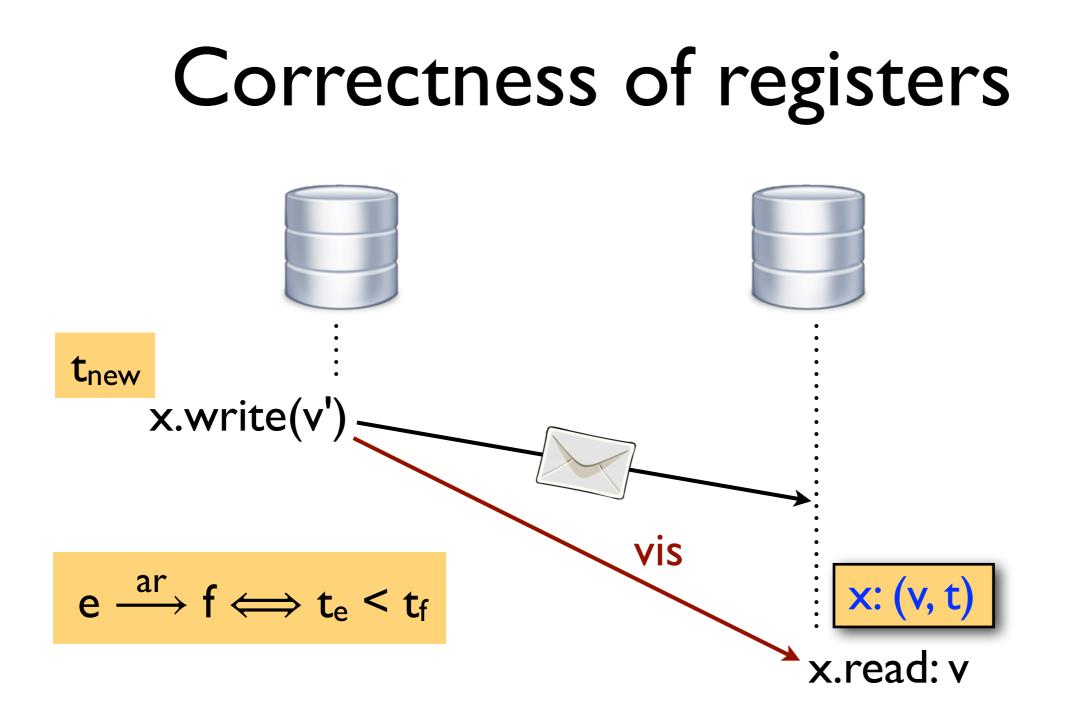


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= the last write in arbitration out of the ones visible to the read, QED.

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- Use the invariants to prove that return values of operations correspond to data type specs

In-between eventual and strong consistency

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Consistency zoo

Eventual consistency

Session guarantees

Causal consistency

Prefix consistency

Sequential consistency

Consistency zoo

Eventual consistency

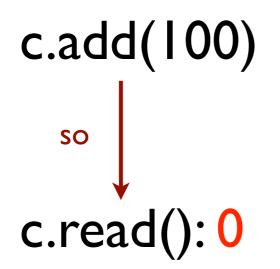
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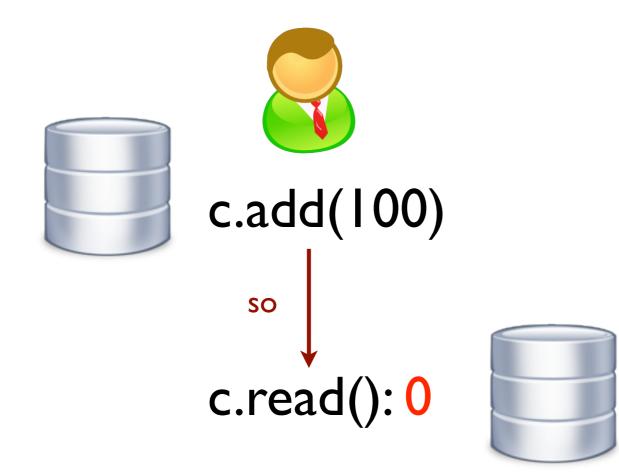
Causal consistency

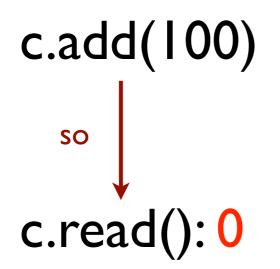
Prefix consistency

Sequential consistency

Keep soundness justifications informal: can be shown using previous techniques

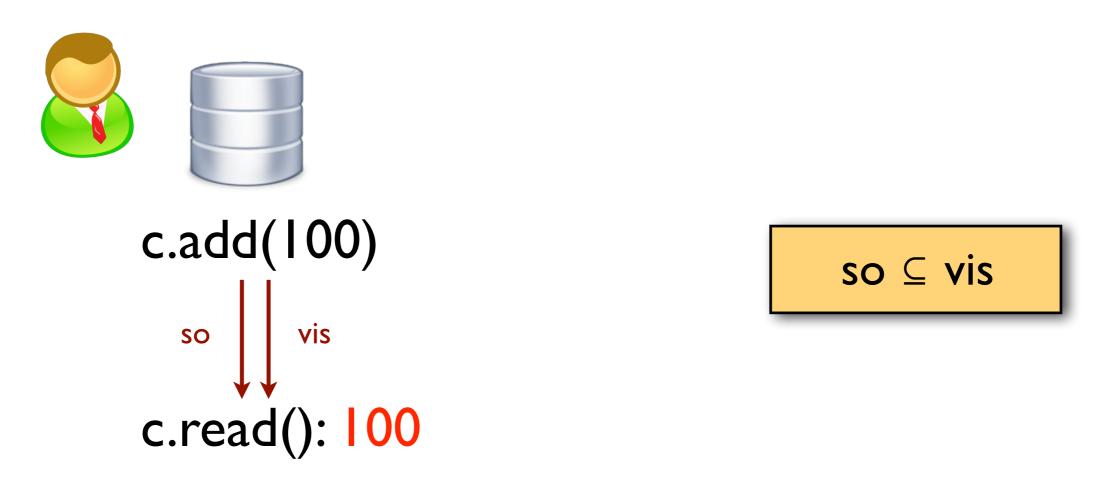






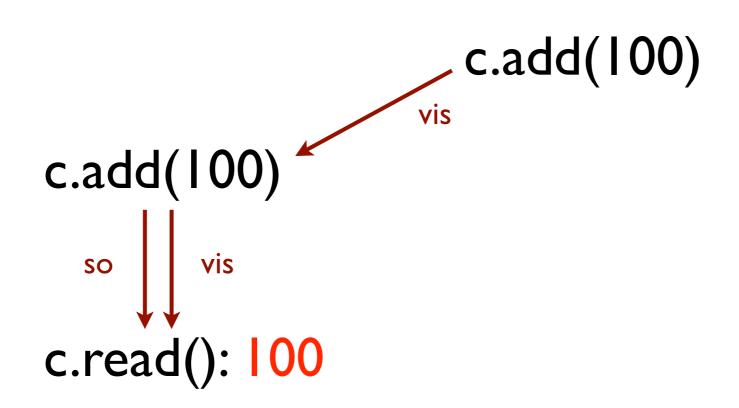


- An operation sees all prior operations by the same process
- Session guarantees: clients only accumulate information

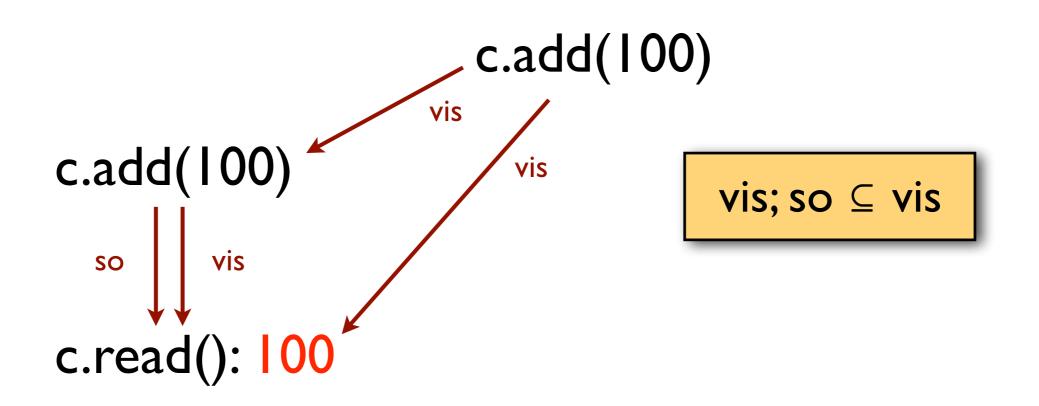


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Monotonic Reads

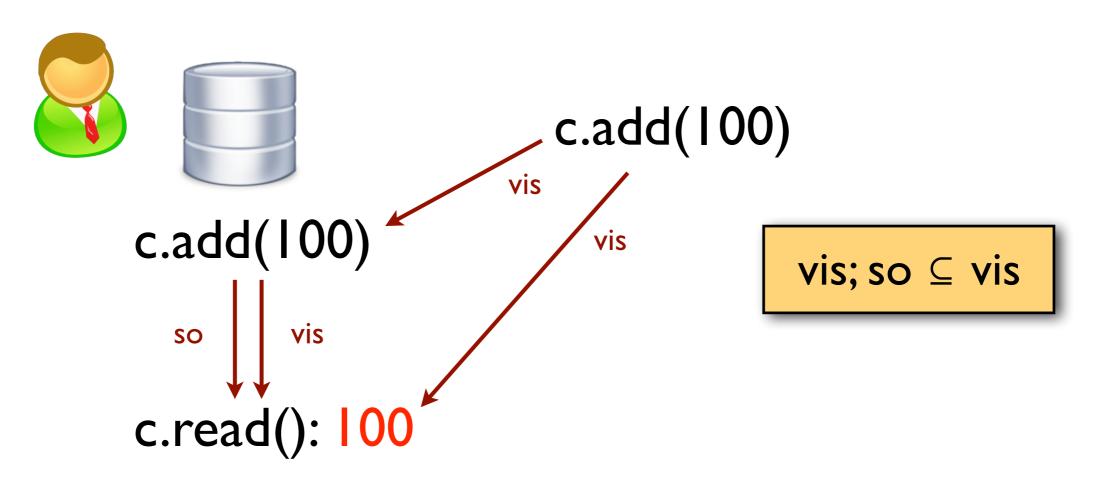


Monotonic Reads

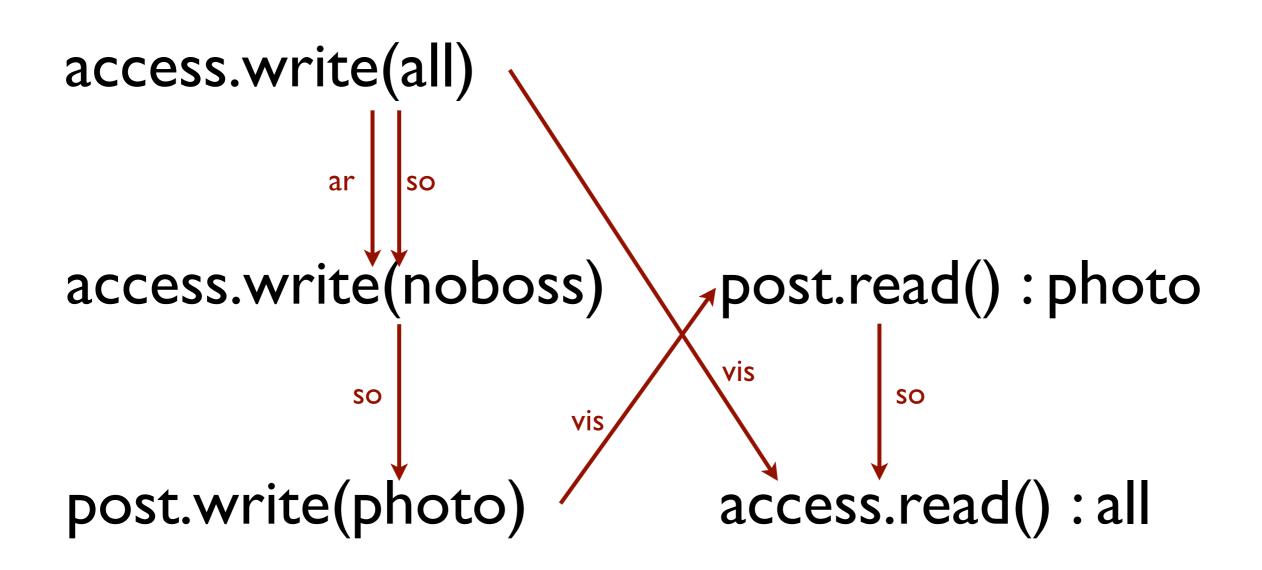


• An operation sees what prior operations by the same session see

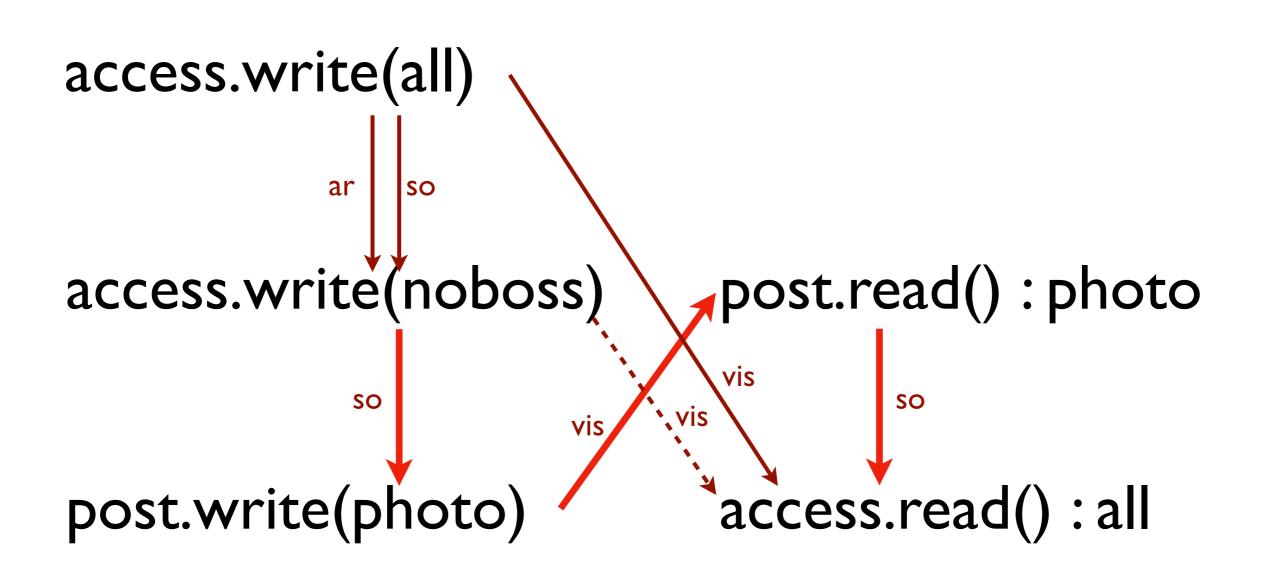
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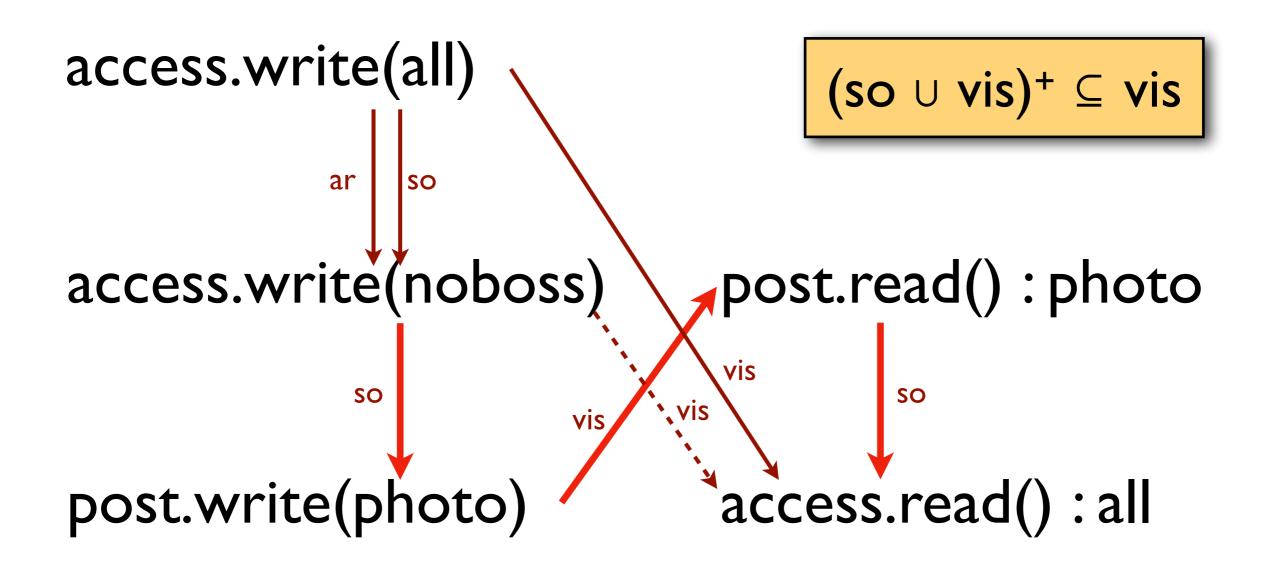


Disallows causality violation anomaly



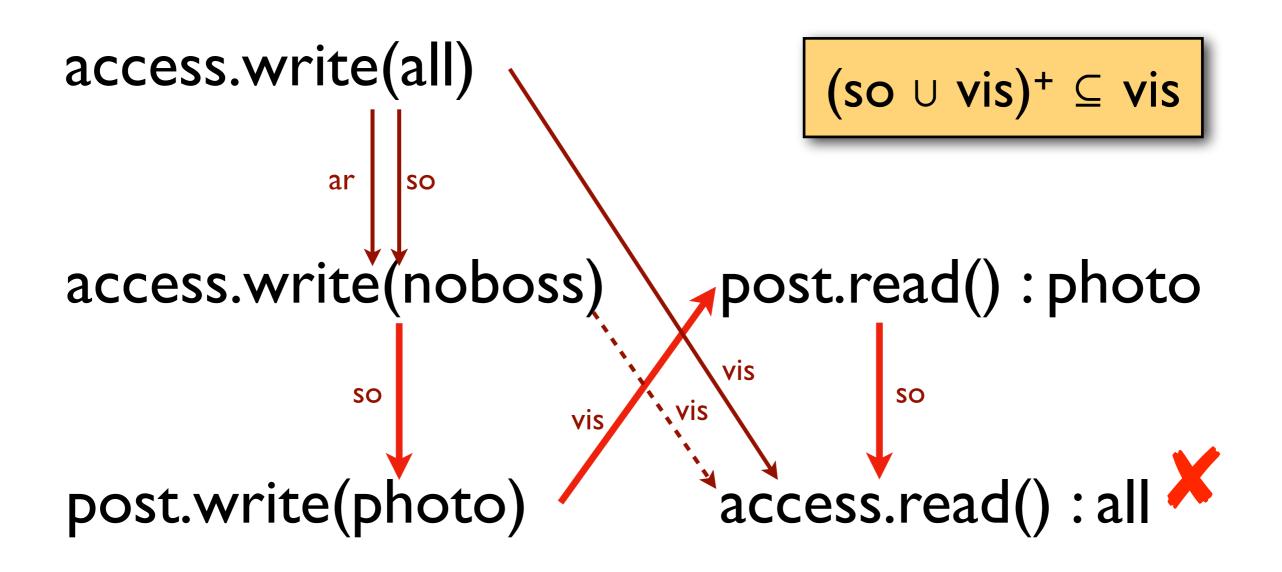
Unintuitive: chain of **so** and **vis** edges from write(noboss) to the read: write happened before the read

Mandate that all actions that happened before an action be visible to it



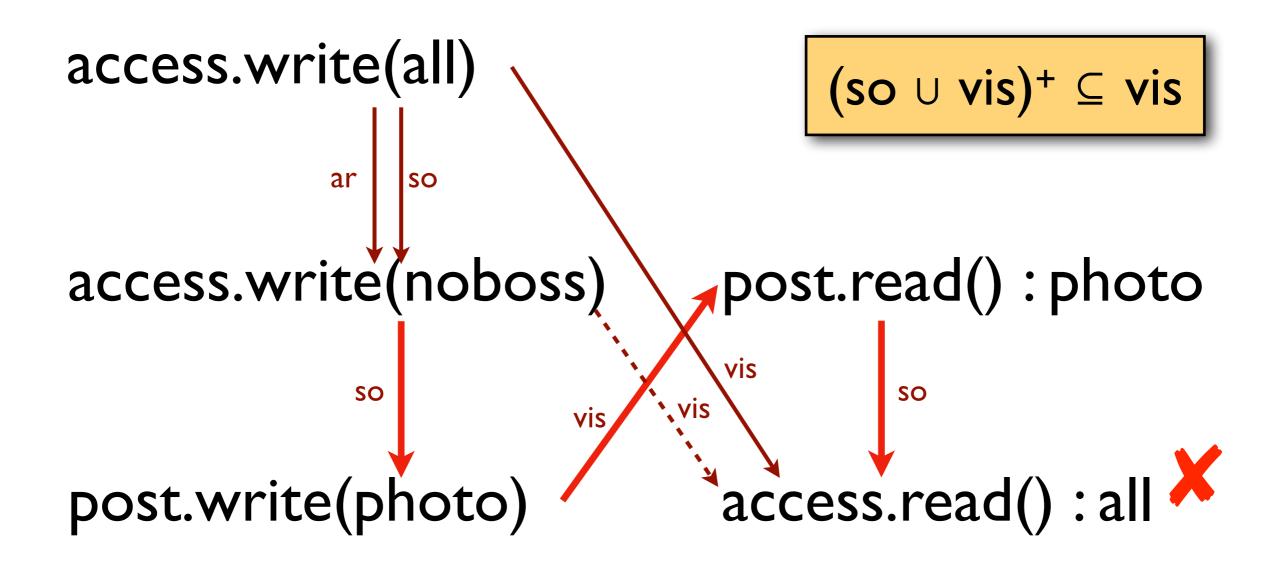
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Implies session guarantees: so \subseteq vis and vis; so \subseteq vis



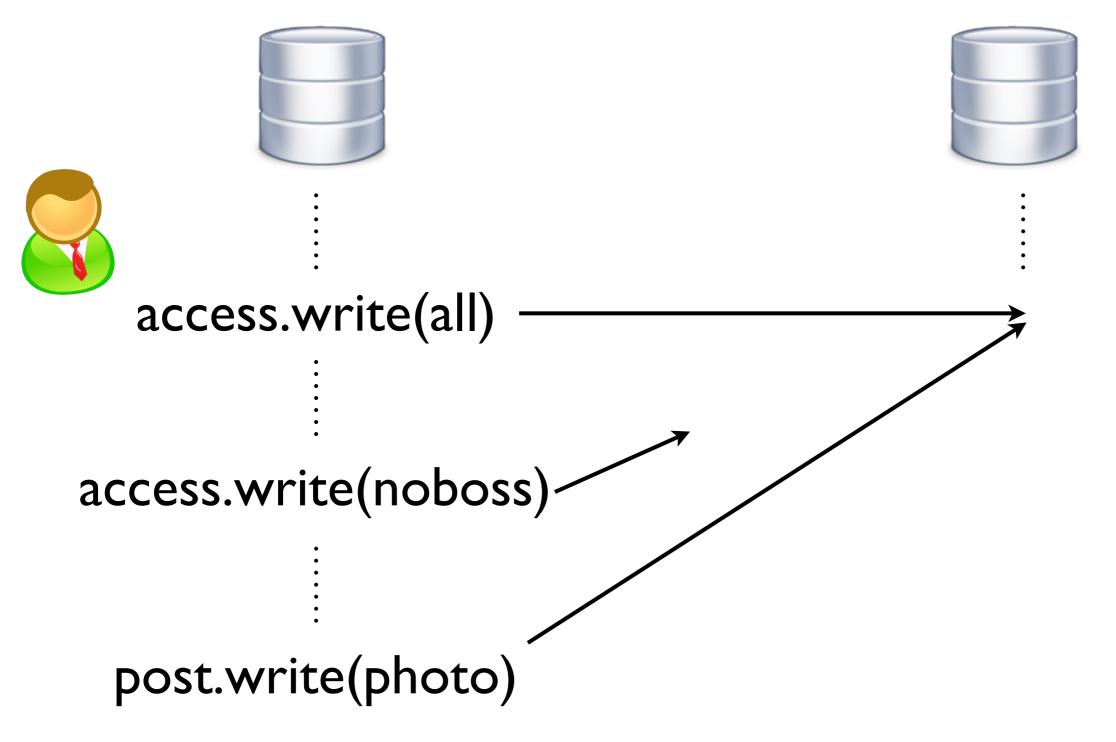




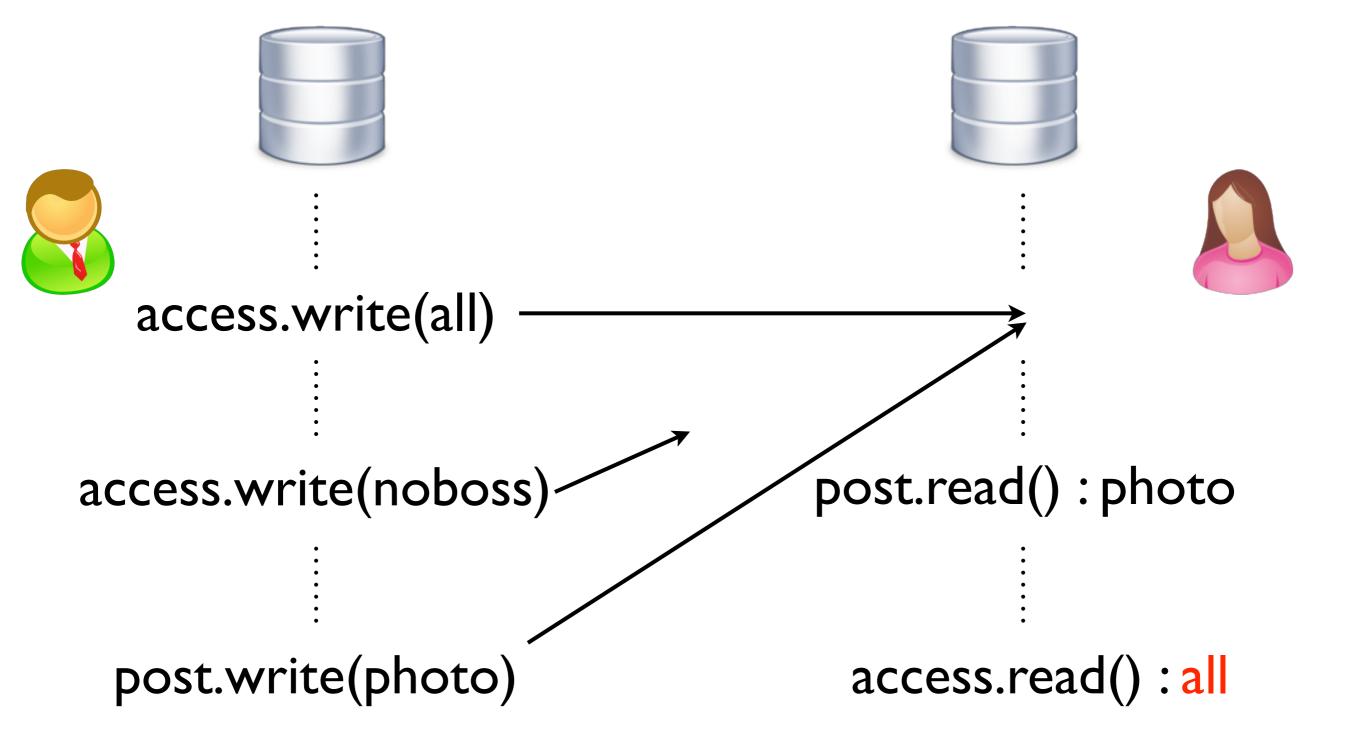
access.write(noboss)

post.write(photo)

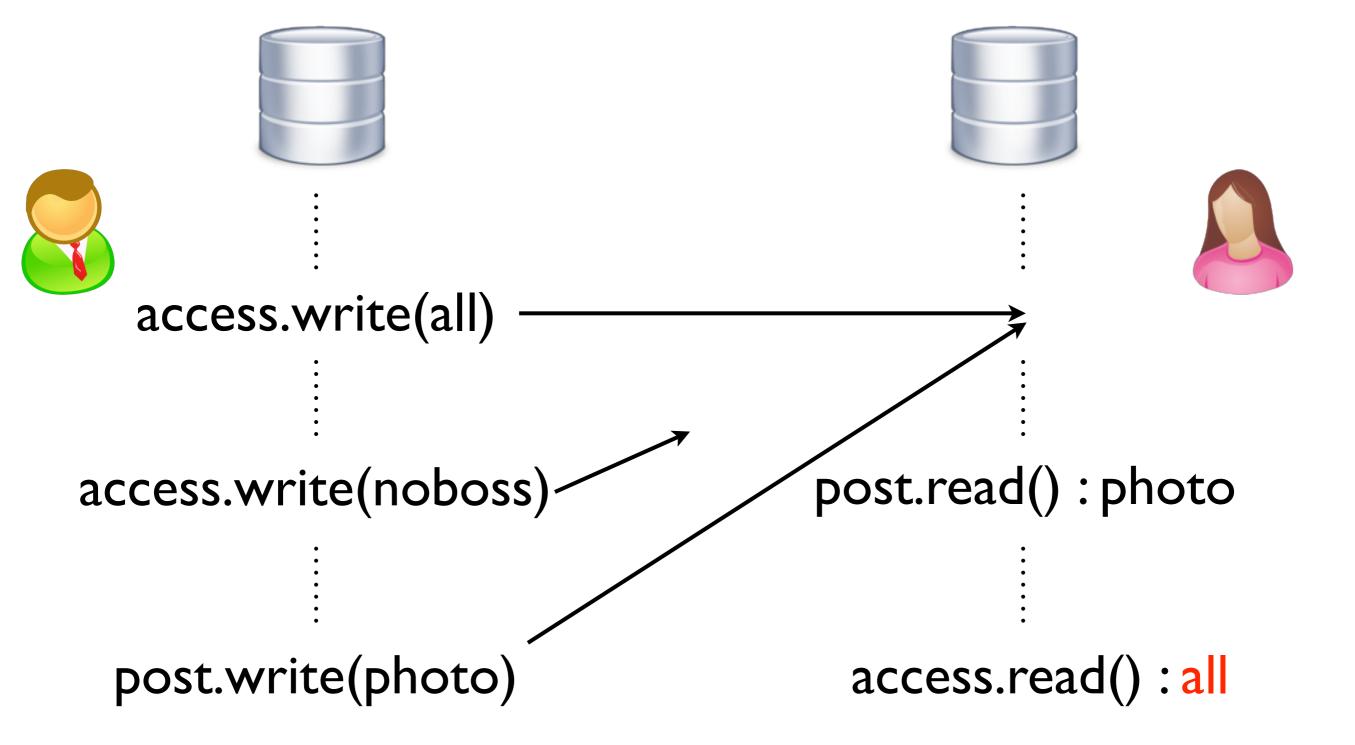
Clients stick to the same replica



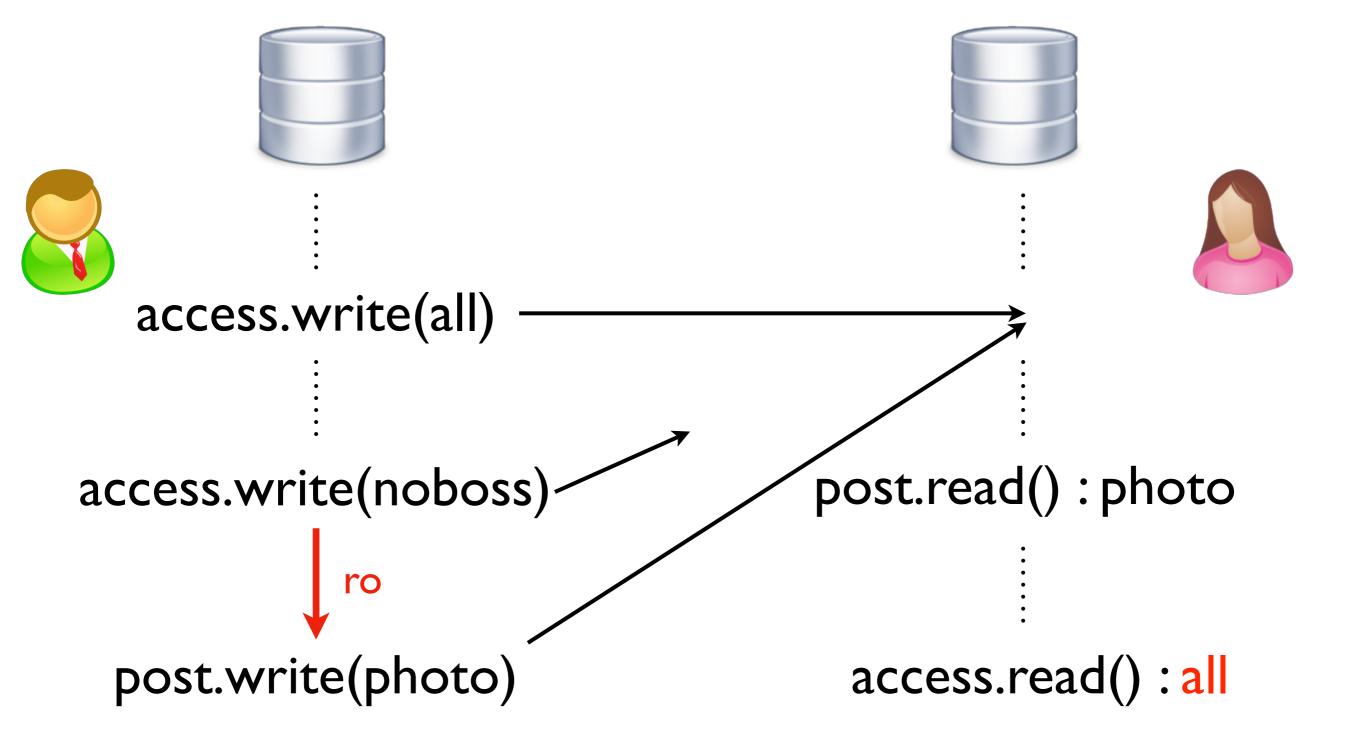
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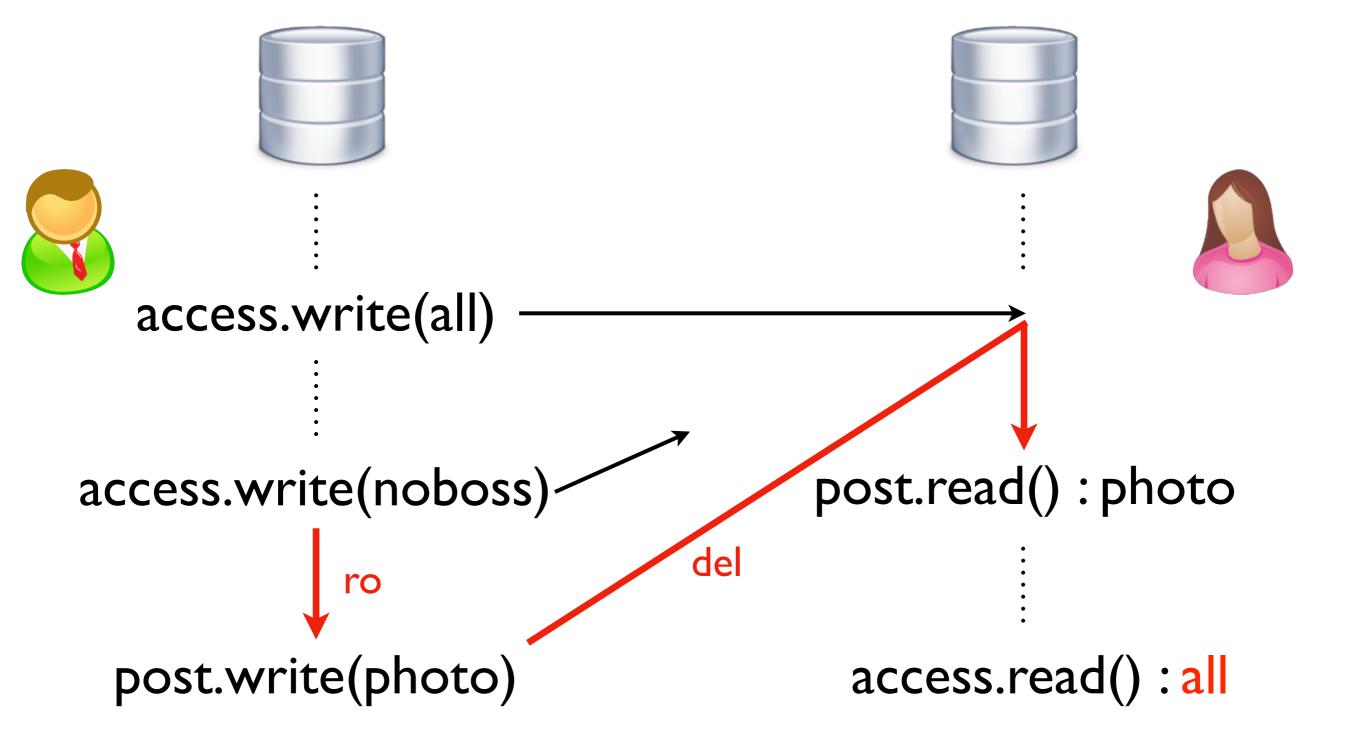
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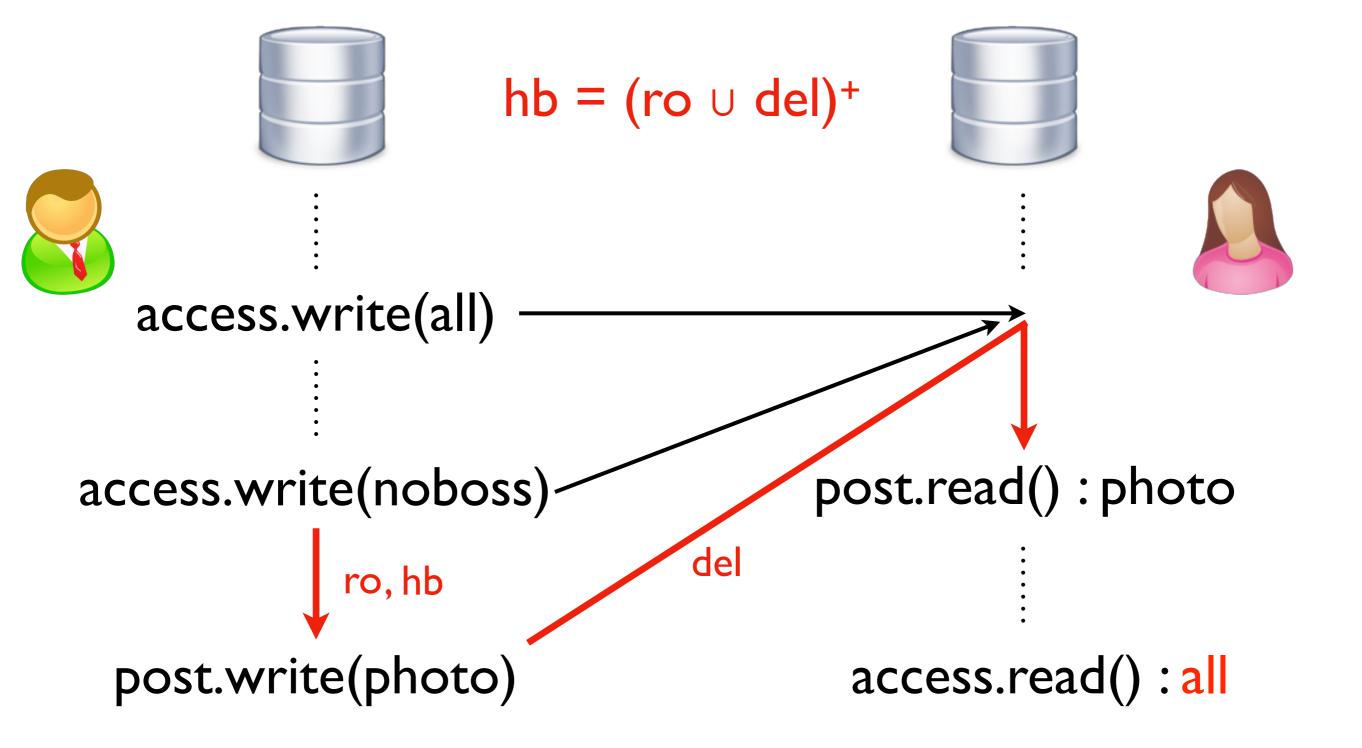
Cannot deliver an operation before delivering its causal dependencies



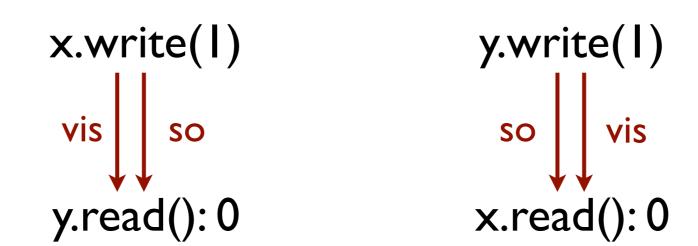
Replica order ro: the order in which operations are issued at a replica

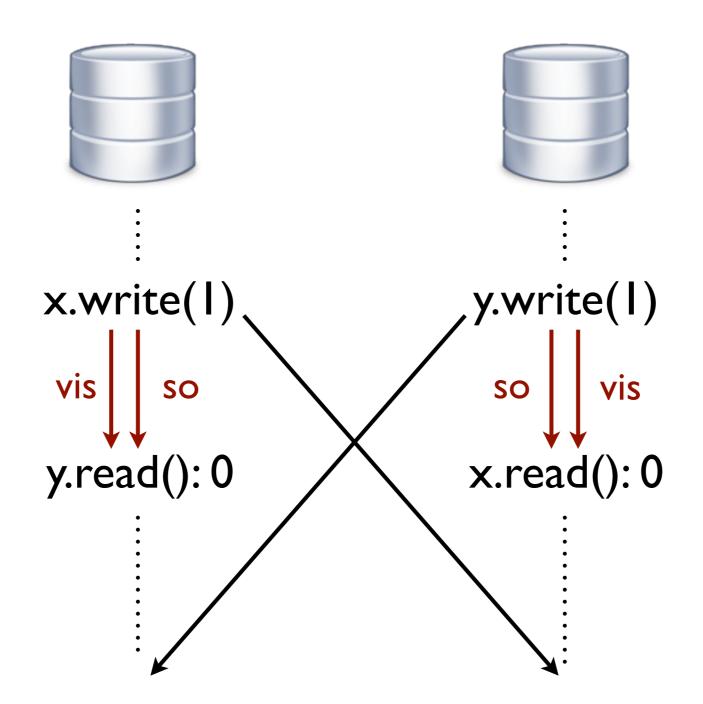


Delivery order del: one operation got delivered before another was issued

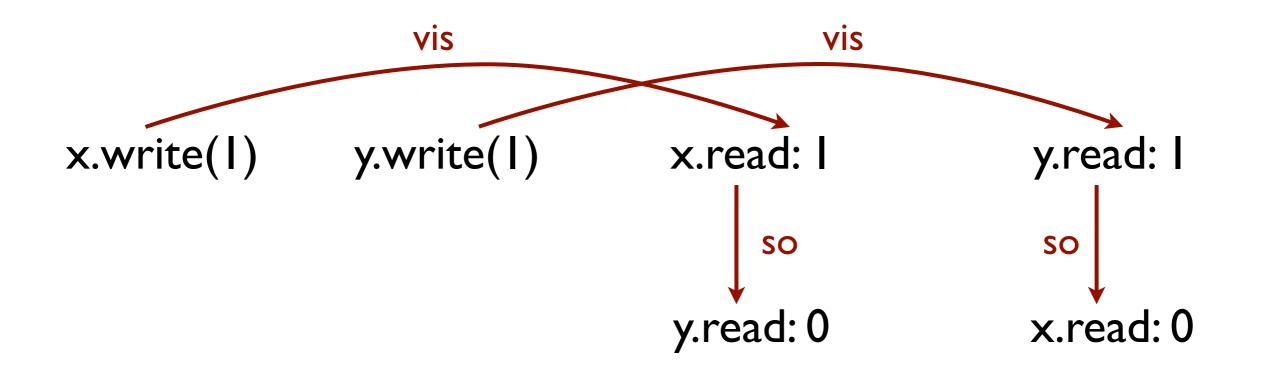


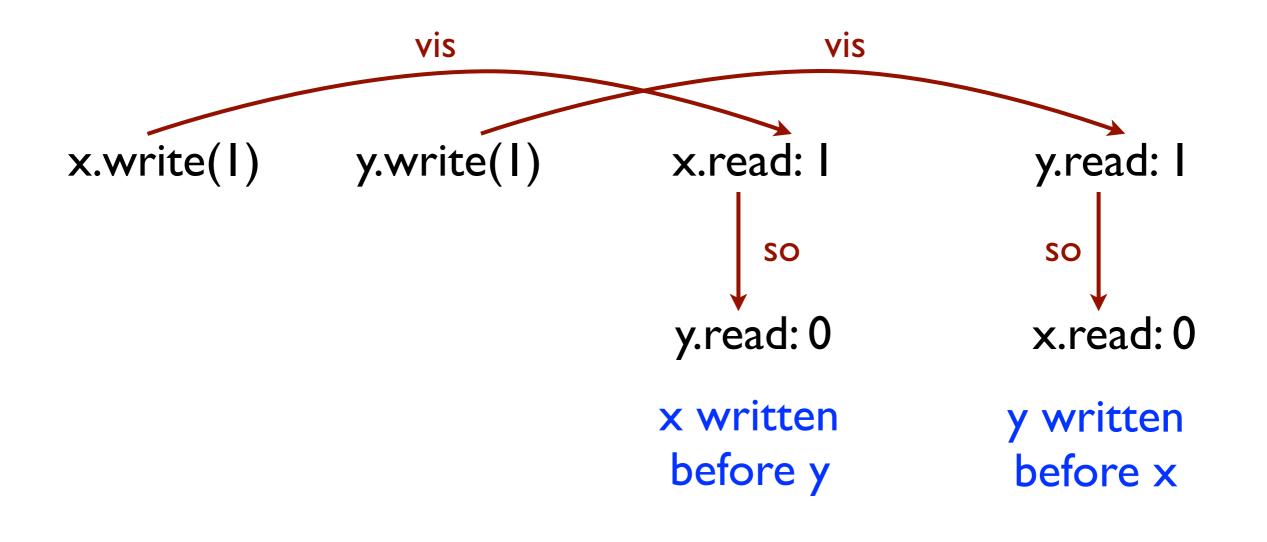
- Causal dependencies of e: hb⁻¹(e)
- An op can only be delivered after all its causal dependencies
- Implementations summarise dependencies concisely

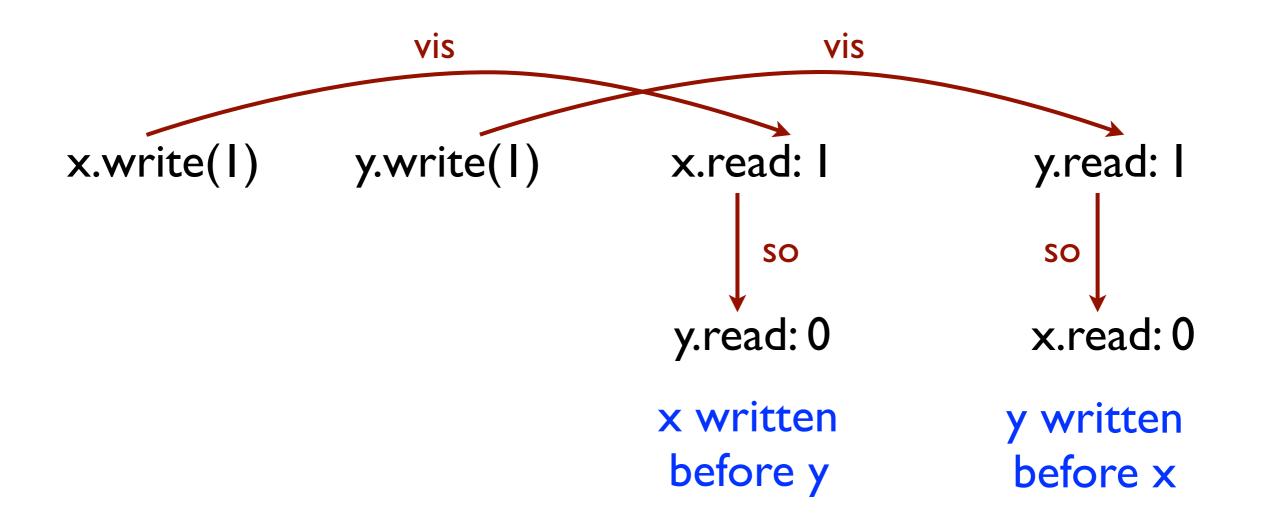




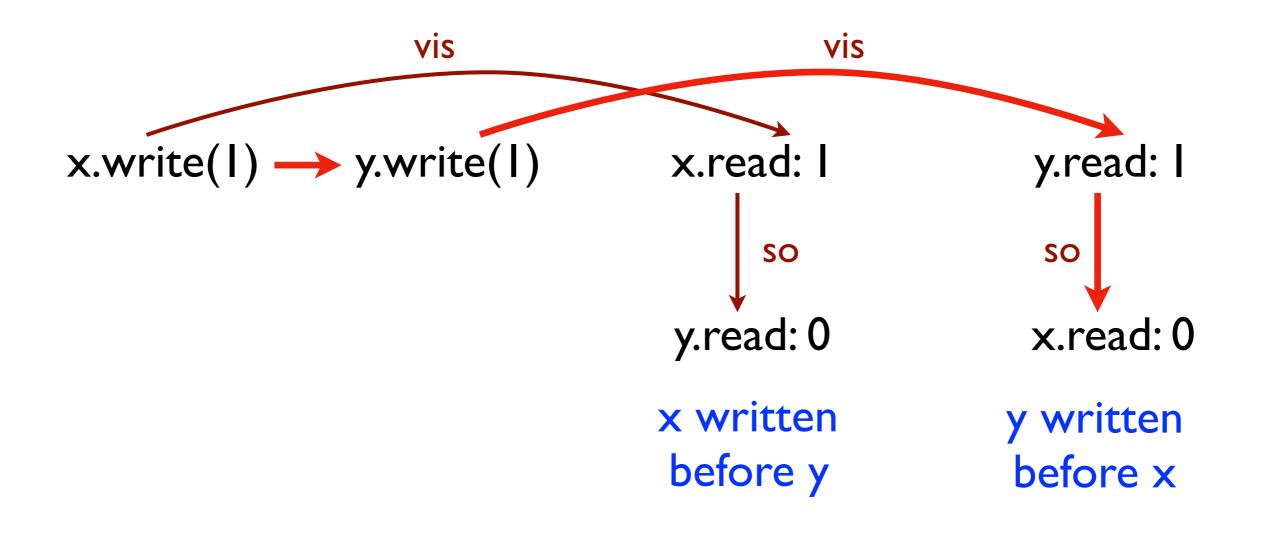
Implementations: updates delivered later

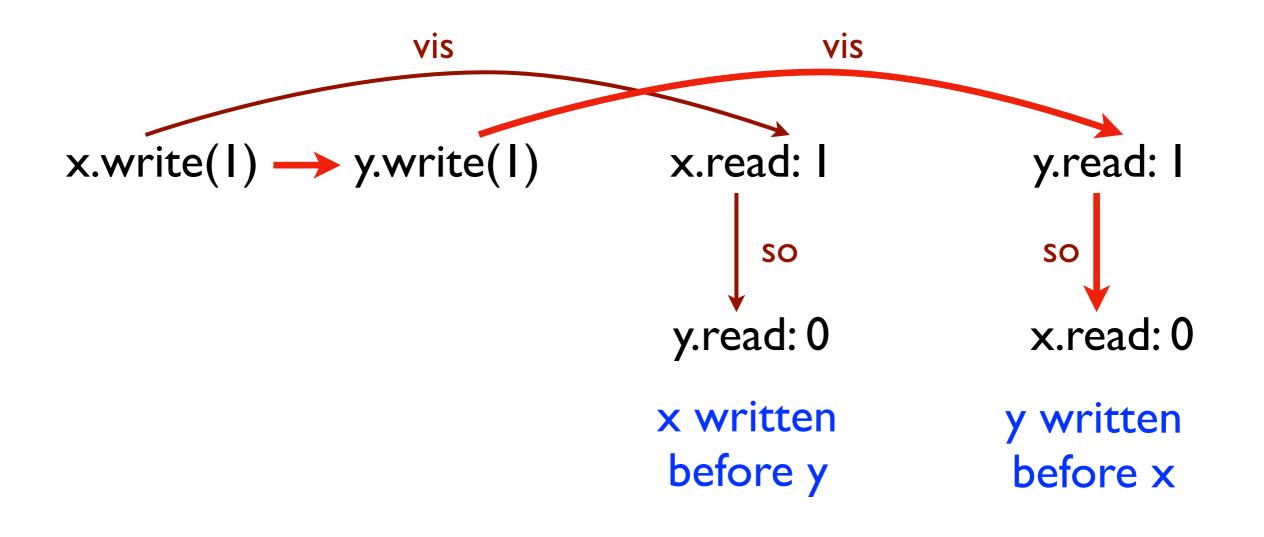




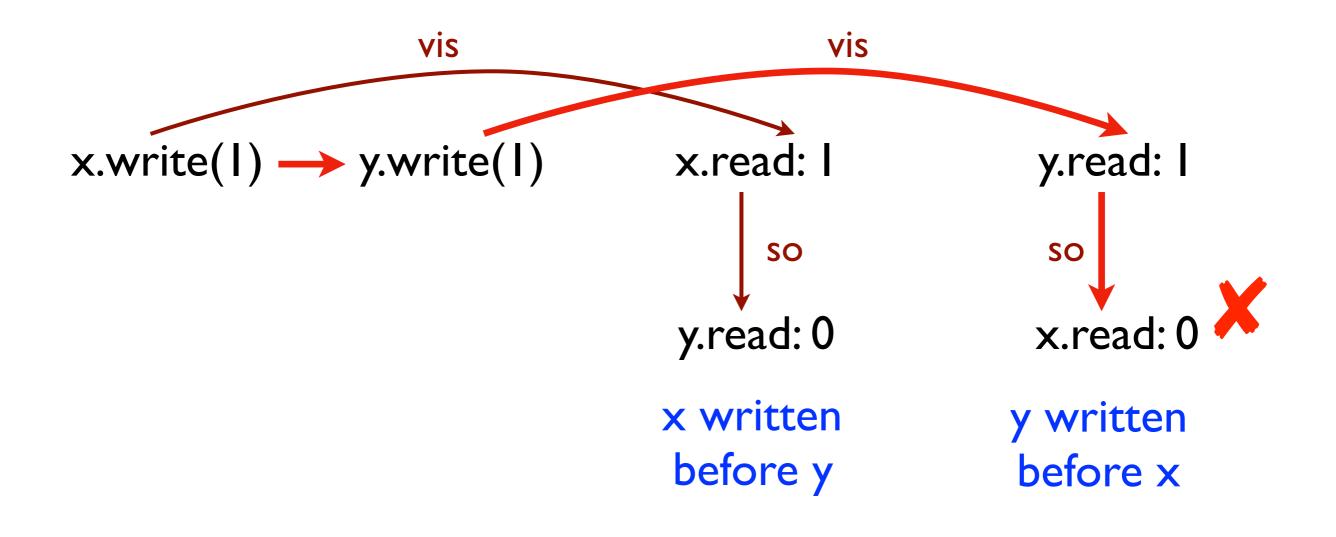


Implementations: no causal dependency between the two writes can be delivered in different orders at different replicas





Not sequentially consistent

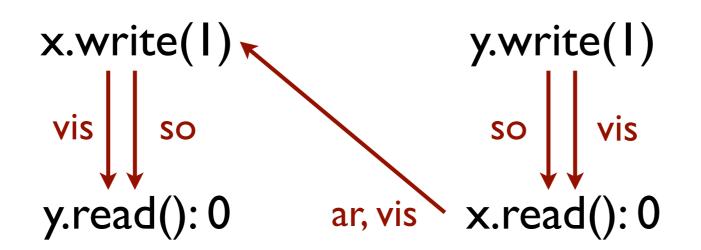


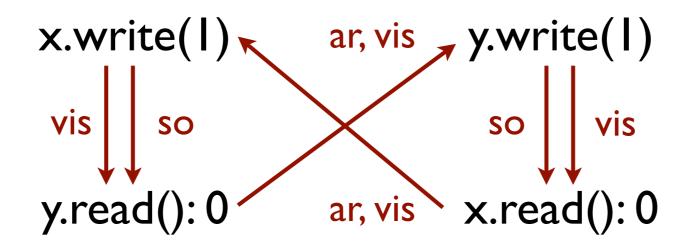
Not sequentially consistent

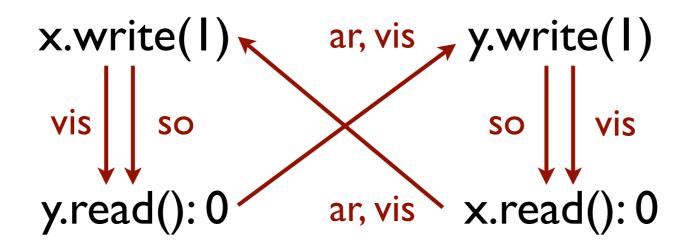
Sequential consistency

- so \subseteq vis and vis is total
- vis \subseteq ar \implies can equivalently require so \subseteq vis = ar
- Every operation sees the effect of all operations preceding it in vis
- Like the original definition with to = vis = ar









No execution with such history

Consistency zoo

- Eventual consistency
- Session guarantees: Dekker, IRIW, causality violation so \subseteq vis, vis; so \subseteq vis
- Causal consistency: Dekker, IRIW
 (so ∪ vis)⁺ ⊆ vis
- Prefix consistency: Dekker
 ar; (vis \ so) ⊆ vis
- Sequential consistency

vis = ar

Shared-memory models



- Sequential consistency first proposed in the context of shared memory (1979)
- Processors and languages don't provide sequential consistency: weak memory models, due to processor and compiler optimisations
- Our specifications similar to weak memory model definitions
- Consistency axioms for last-writer-wins registers
 ~ shared-memory models

Consistency zoo

- Eventual consistency
- Session guarantees: Dekker, IRIW, causality violation so \subseteq vis, vis; so \subseteq vis
- Causal consistency: Dekker, IRIW
 (so ∪ vis)⁺ ⊆ vis
- Prefix consistency: Dekker ar; (vis \ so) ⊆ vis
- Sequential consistency

vis = ar

for last-writer-wins = C++ release/acquire

- Eventual consistency
- Session guarantees so \subseteq vis, vis; so \subseteq vis
- Causal consistency $(so \cup vis)^+ \subseteq vis$
- Prefix consistency ar; (vis $\ so$) \subseteq vis
- Sequential consistency

vis = ar

- What's the best we can do while staying available under network partitionings?
- Causal consistency is a strongest such model [Attiya et al., 2015]

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- Sequential consistency
 vis = ar

- What's the best we can do while staying available under network partitionings?
- Causal consistency is a strongest such model [Attiya et al., 2015]

Terms and conditions apply:

- for a certain version of CC and a certain class of implementations
- a strongest model: cannot be strengthened, but can be other alternative incomparable models

- Application of eventual consistency collaborative editing: Google Docs, Office Online
- At the core: list data type (of formatted characters)
- List data type has an inherently high metadata overhead: can't discard a character when deleting it from a Google Docs document! [Attiya et al., 2016]
- Discarding may allow previously deleted elements to reappear

Determining the right level of consistency

Application correctness

 Does an application satisfy a particular correctness property?

Integrity invariants: account balance is non-negative

Is an application robust against a particular consistency model?

Application behaves the same as when using a strongly consistent database

Application correctness

 Does an application satisfy a particular correctness property?

Integrity invariants: account balance is non-negative

Is an application robust against a particular consistency model?

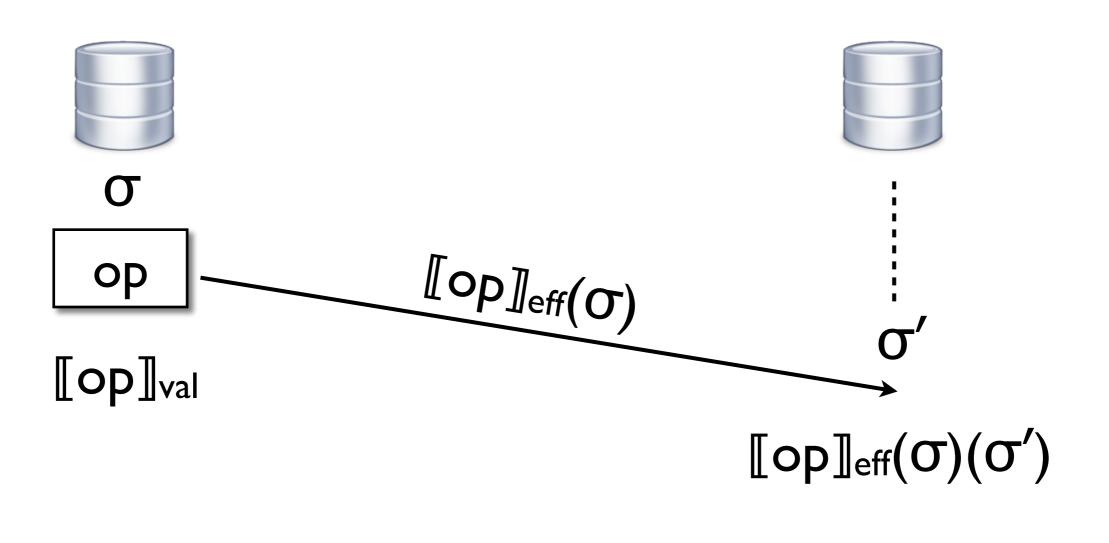
Application behaves the same as when using a strongly consistent database

Challenge

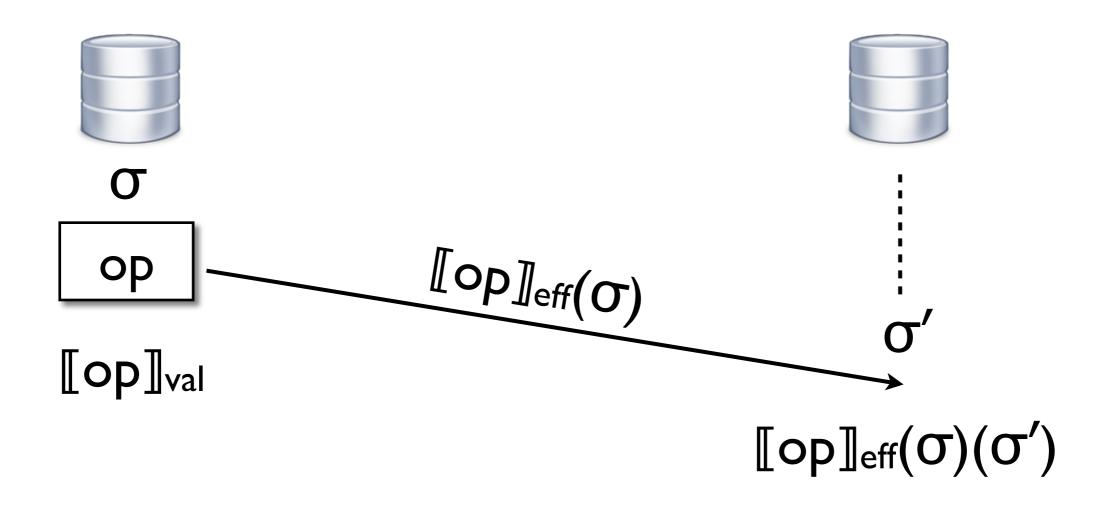
Vanilla weak consistency often too weak to preserve correctness

Need to strengthen consistency in parts of the application

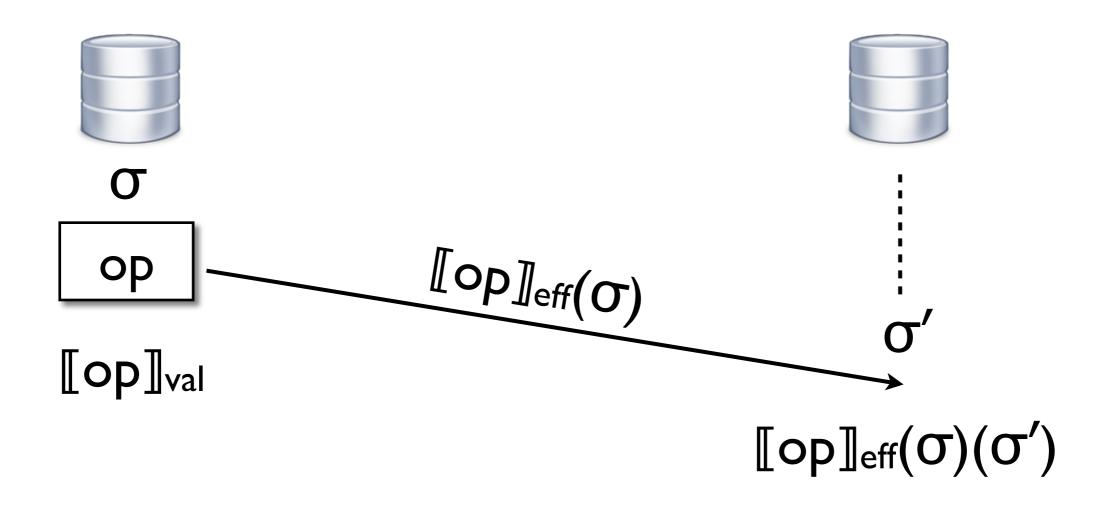
Deposits



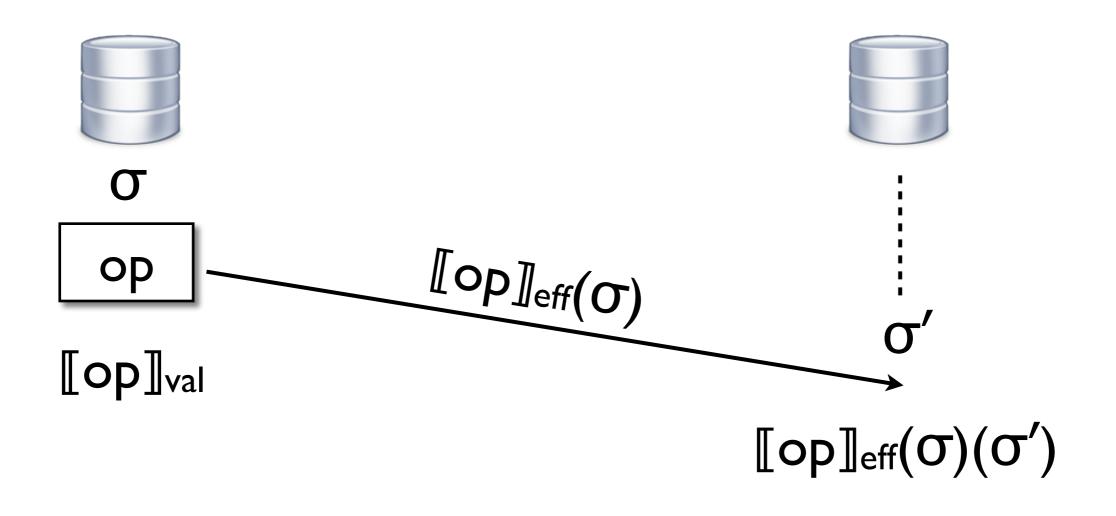
 $[add(100)]_{eff}(\sigma) = \lambda\sigma'.(\sigma' + 100)$



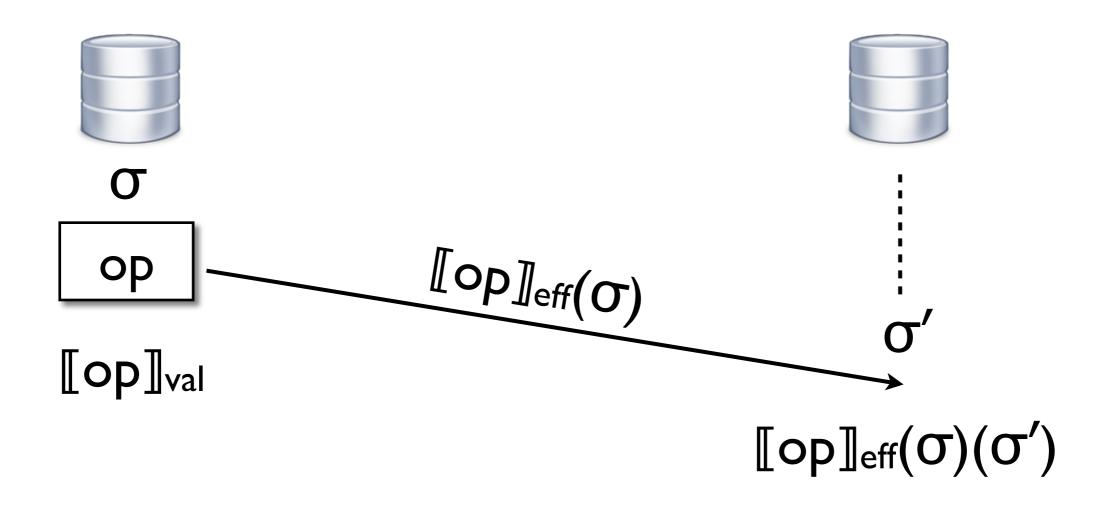
 $[[withdraw(100)]]_{eff}(\sigma) =$ if $\sigma \ge 100$ then $(\lambda \sigma'. \sigma' - 100)$ else $(\lambda \sigma'. \sigma')$



 $[[withdraw(100)]]_{eff}(\sigma) =$ if $\sigma \ge 100$ then $(\lambda \sigma', \sigma' - 100)$ else $(\lambda \sigma', \sigma')$



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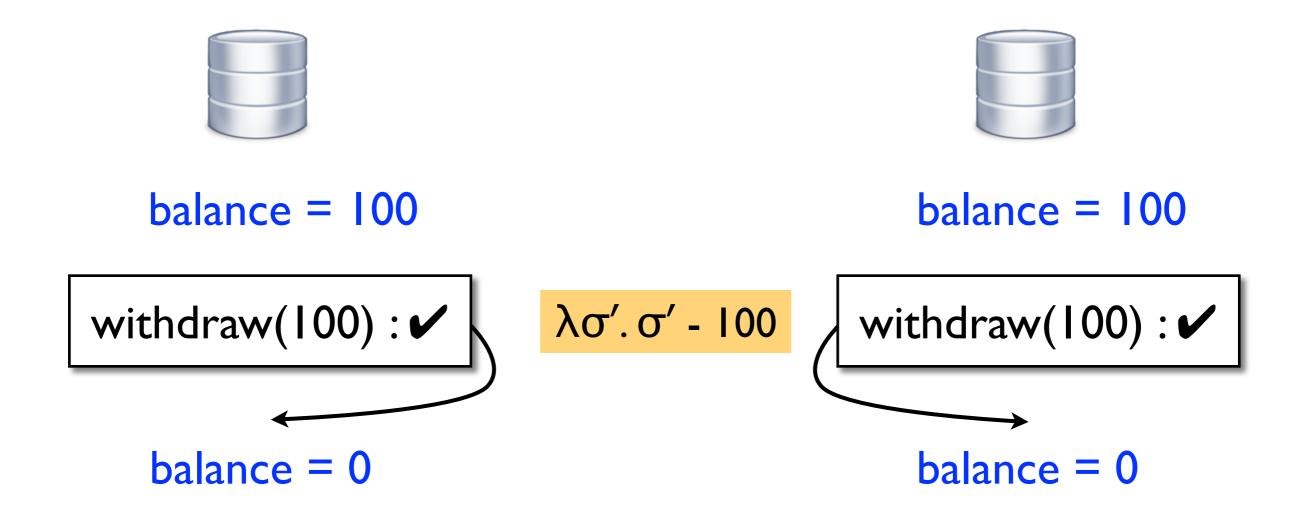
balance = 100

withdraw(100) : 🗸

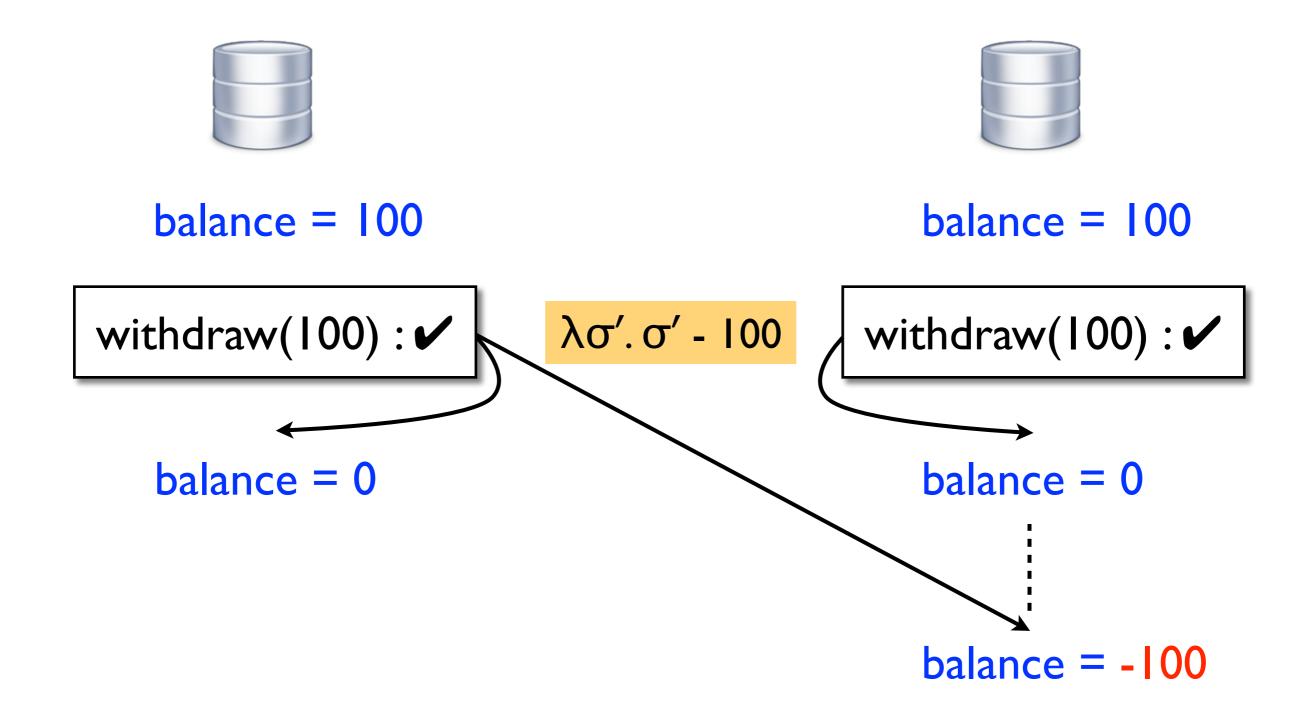


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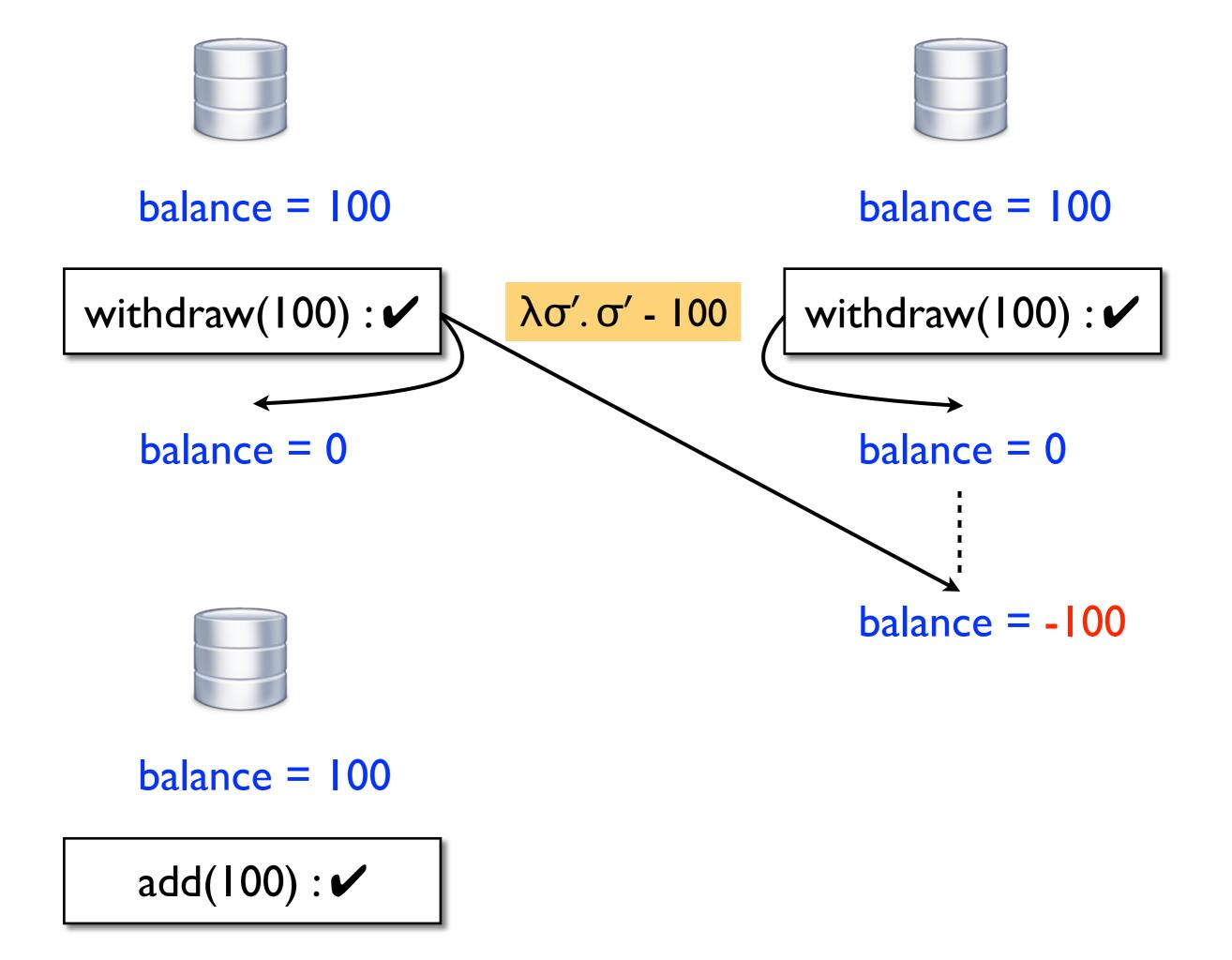
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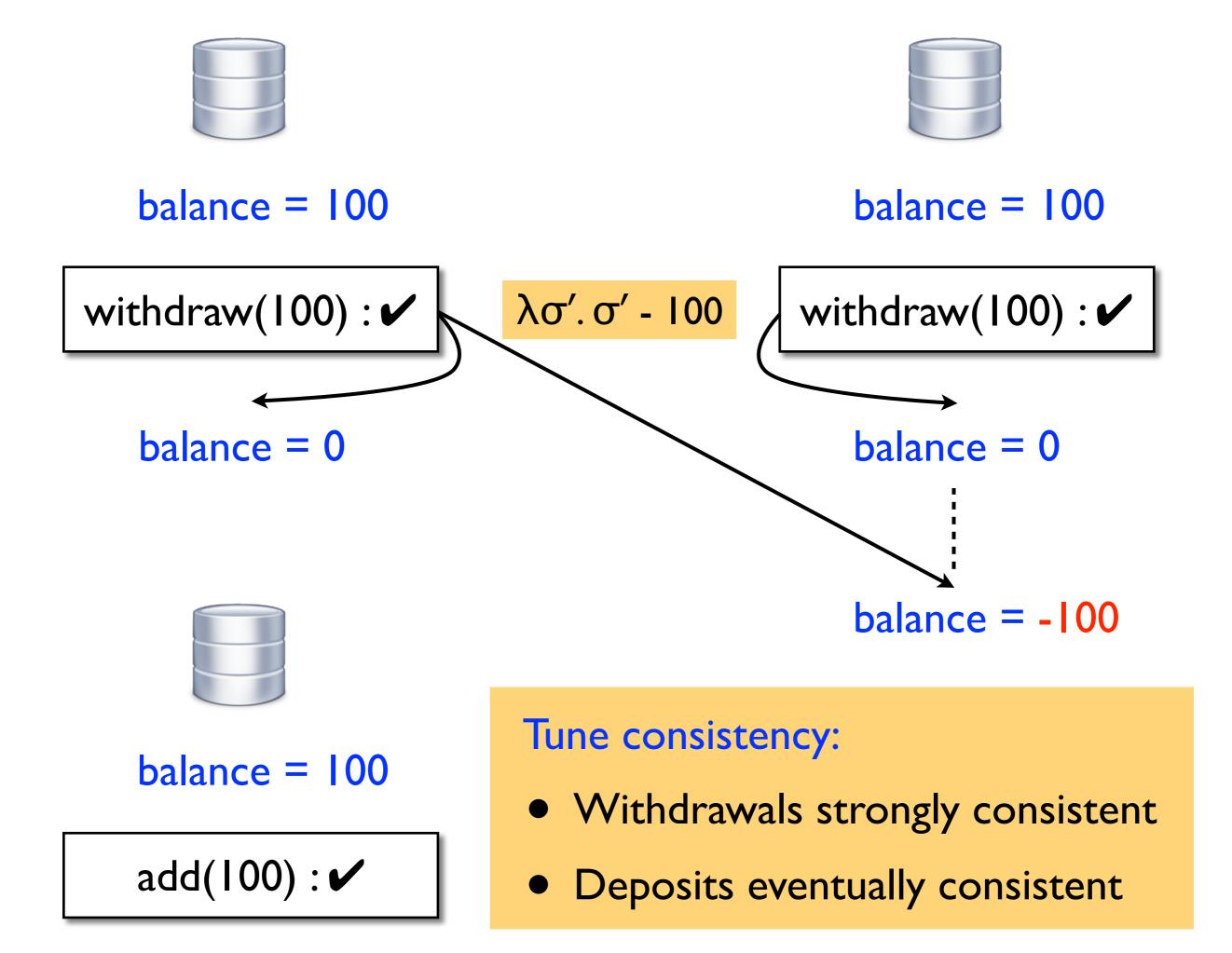


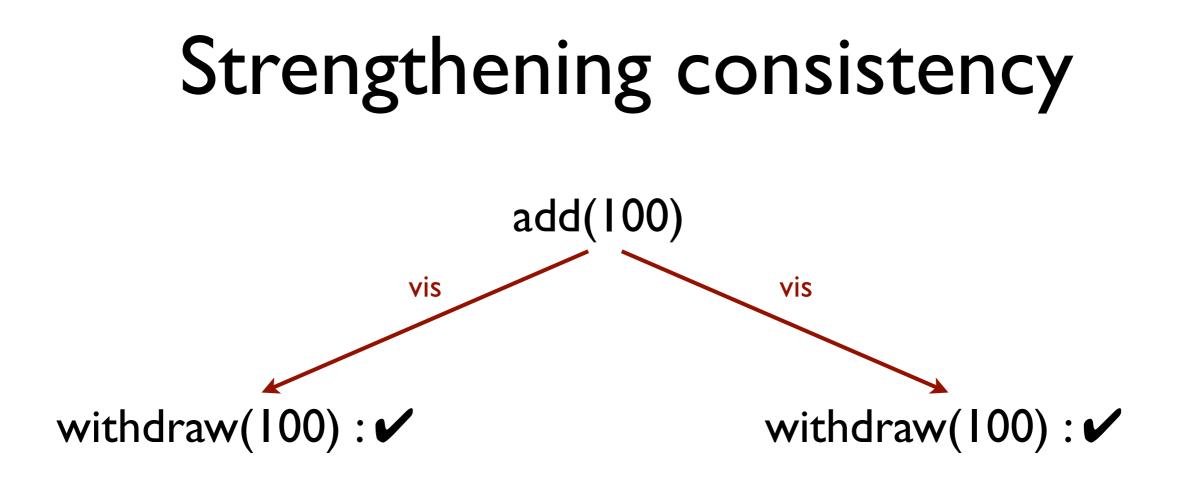
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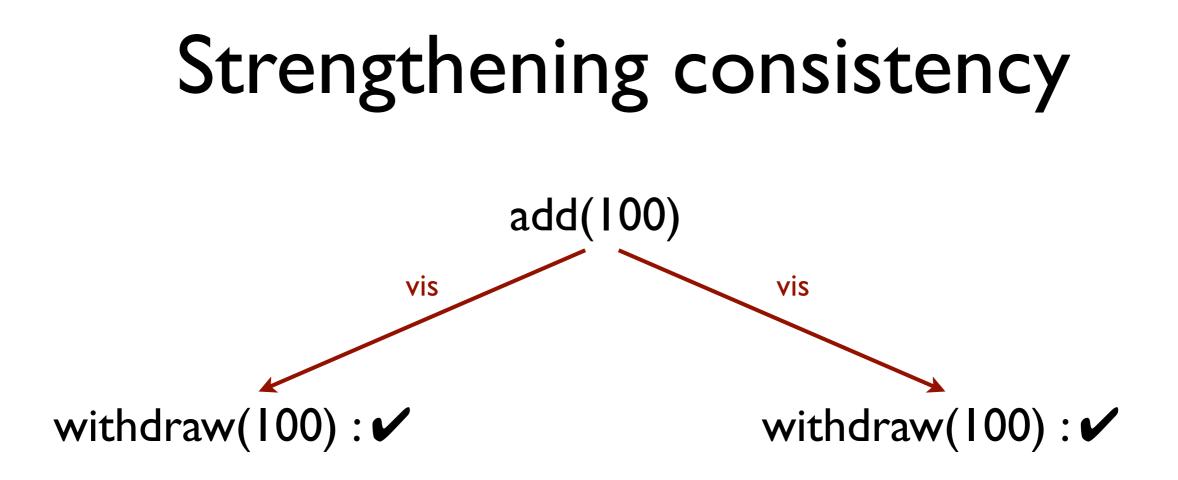
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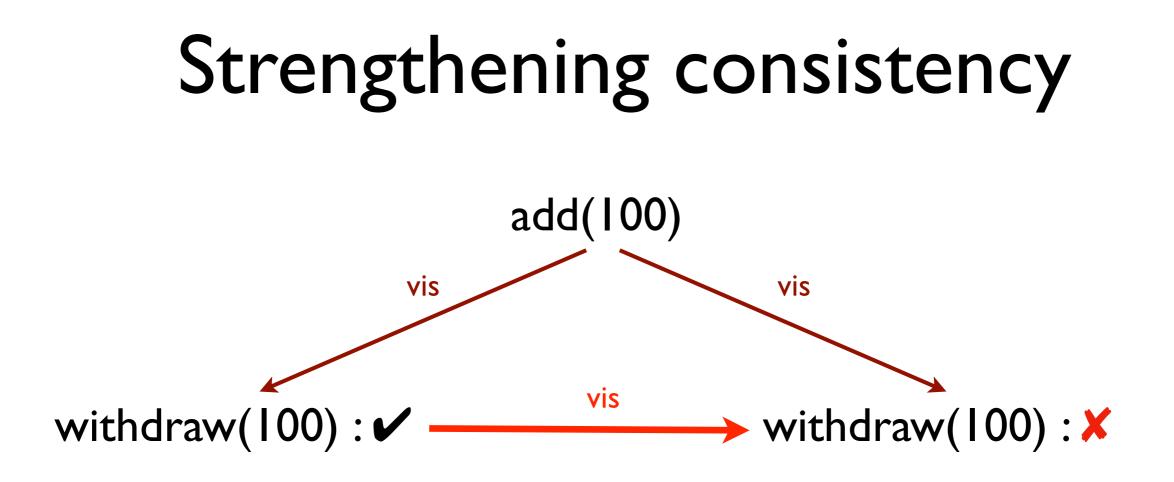




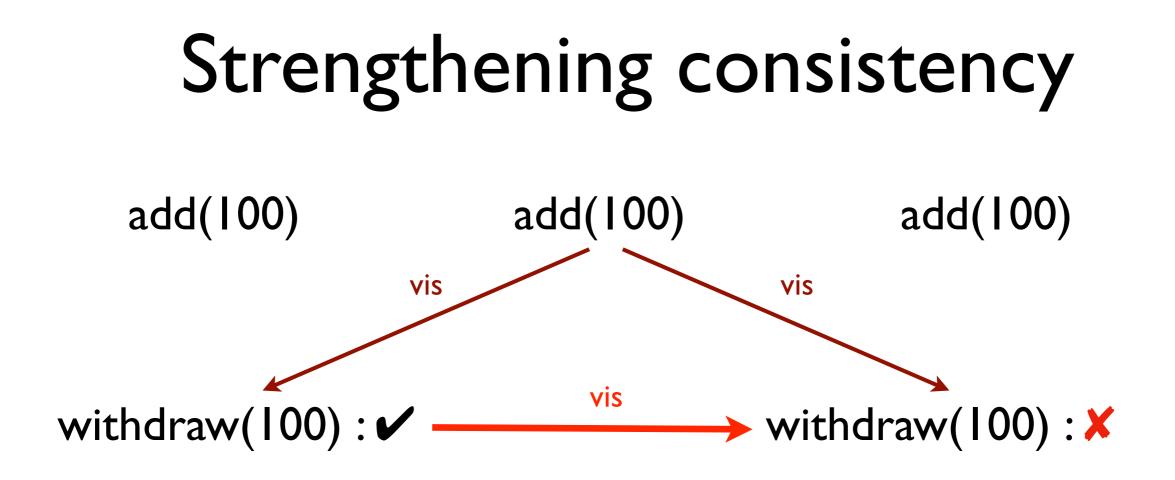
- Baseline model: causal consistency
- Problem: withdrawals are causally independent



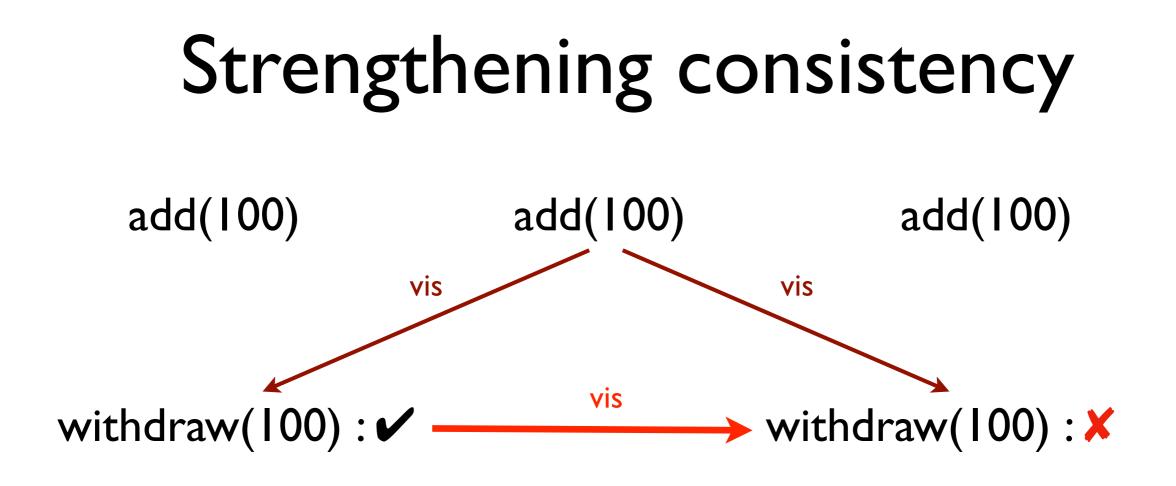
- Symmetric conflict relation on operations: $\bowtie \subseteq \mathsf{Op} \times \mathsf{Op}$, e.g., withdraw \bowtie withdraw
- Conflicting operations cannot be causally independent: $\forall e, f \in E. op(e) \bowtie op(f) \Longrightarrow e \xrightarrow{vis} f \lor f \xrightarrow{vis} e$



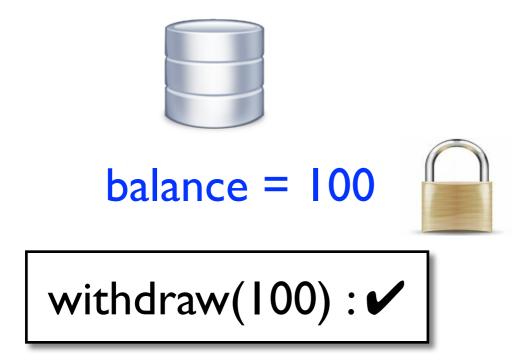
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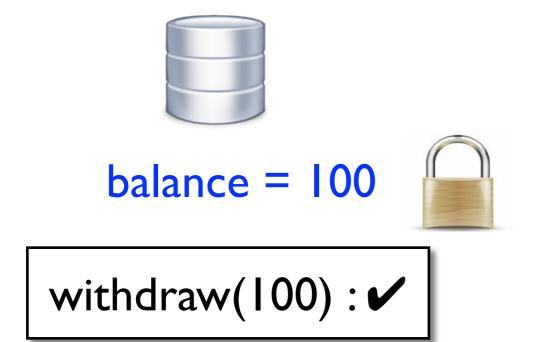


- Symmetric conflict relation on operations: $\bowtie \subseteq \mathsf{Op} \times \mathsf{Op}$, e.g., withdraw \bowtie withdraw
- Conflicting operations cannot be causally independent: $\forall e, f \in E. op(e) \bowtie op(f) \Longrightarrow e \xrightarrow{vis} f \lor f \xrightarrow{vis} e$
- No constraints on additions: ¬(add × op)



- Implementation requires replicas executing withdraw() to synchronise
- add() doesn't need synchronisation







balance = 100

withdraw(100) : ?

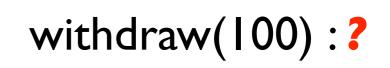
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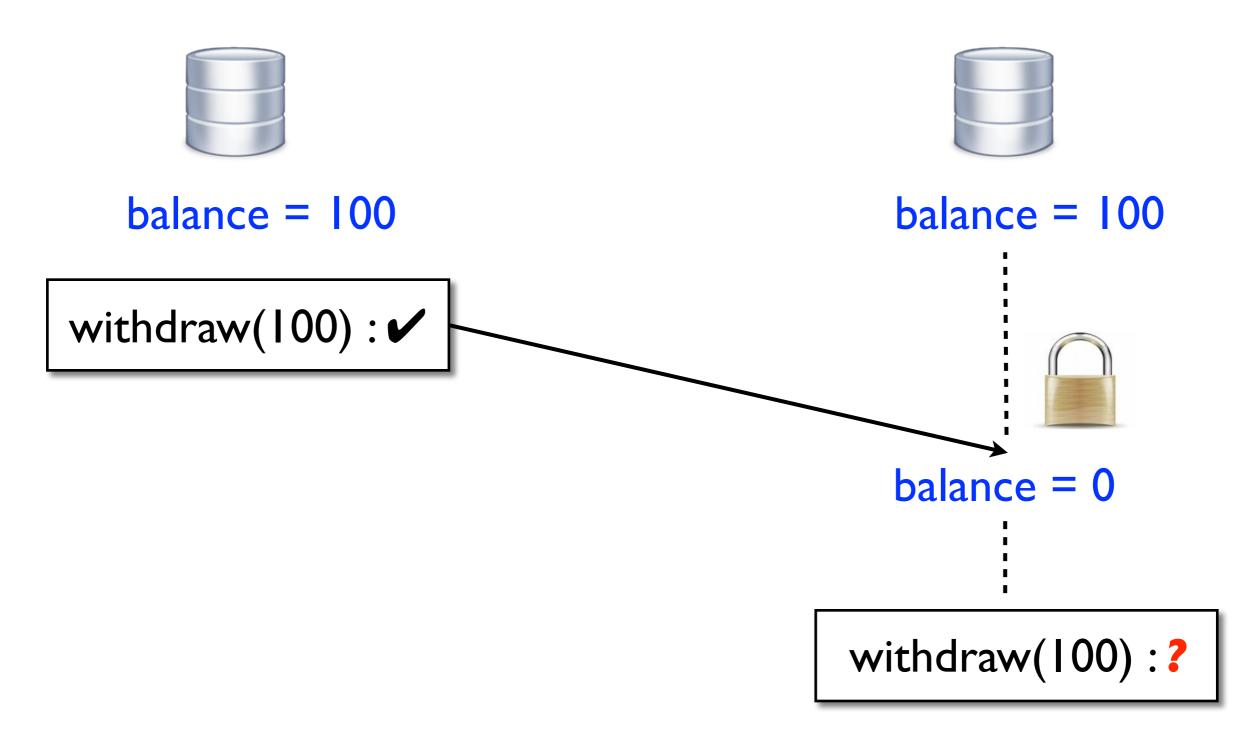
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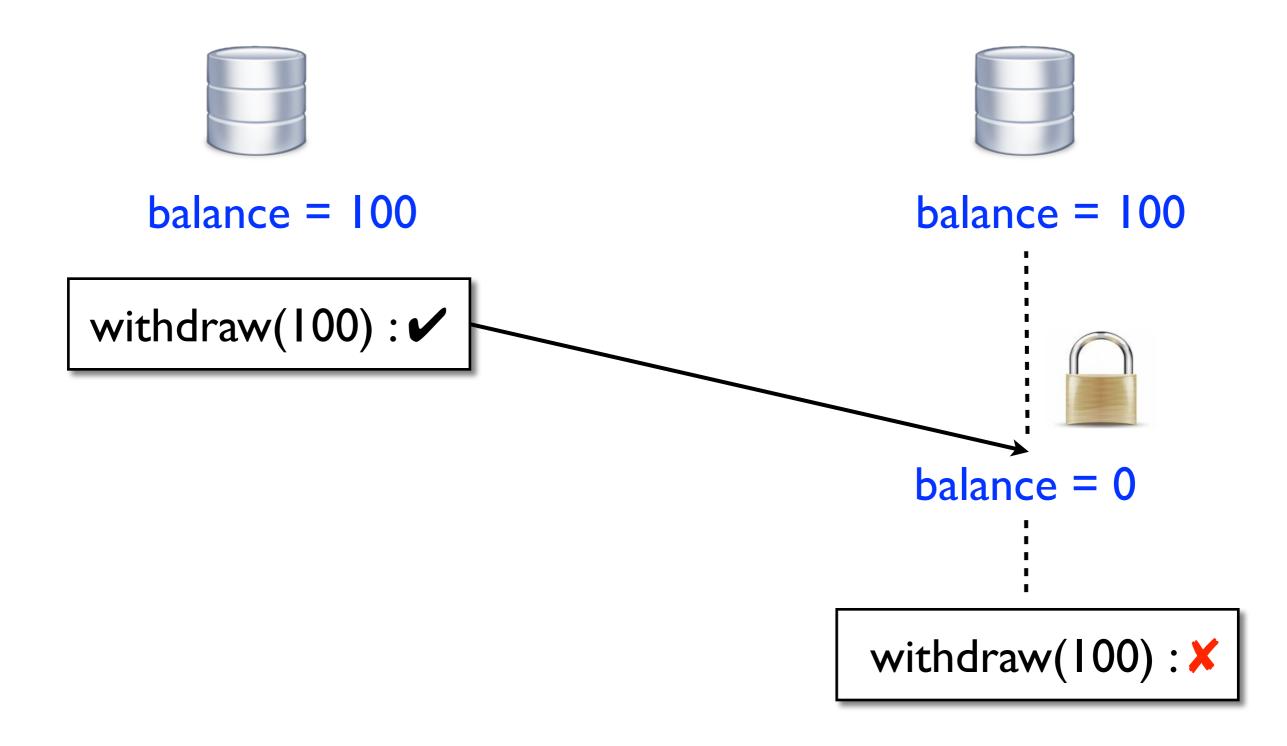
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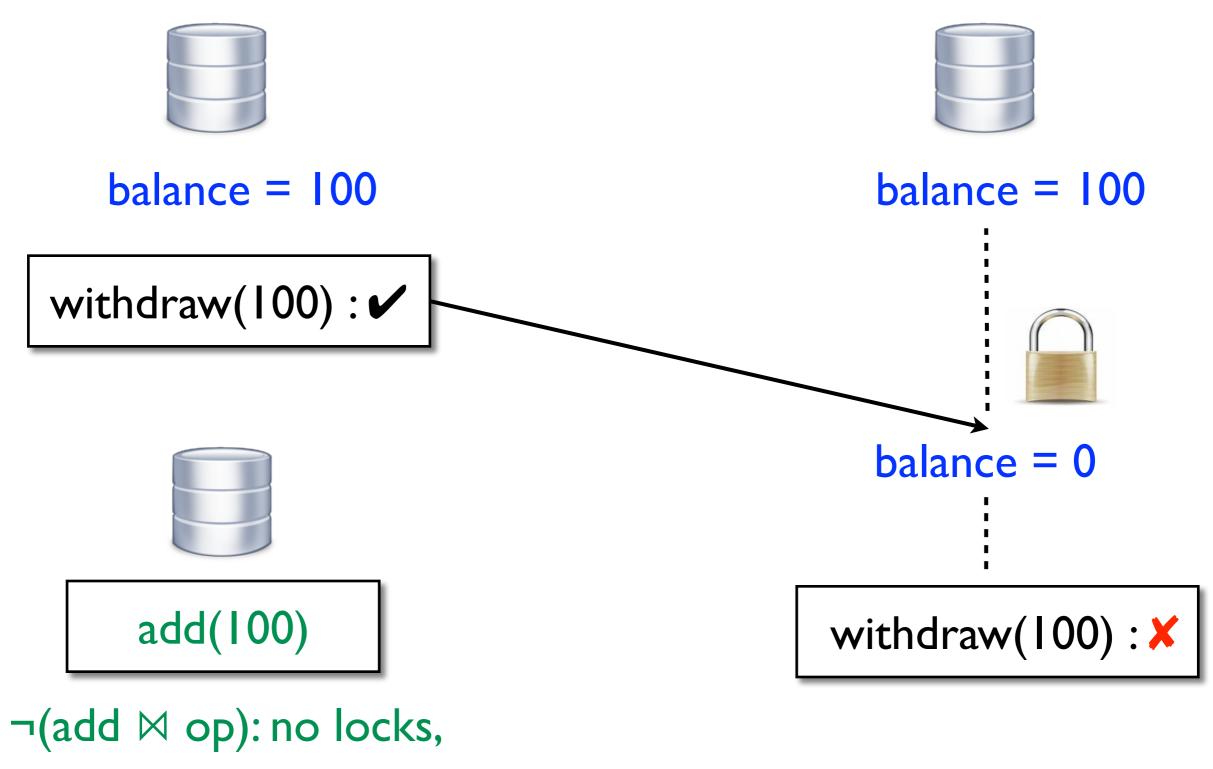






Acquiring the lock requires bringing all operations the replica holding it knows about





so no synchronisation

Consistency choices

- Databases with multiple consistency levels:
 - Commercial: Amazon DynamoDB, Microsoft DocumentDB
 - Research: Li⁺ OSDI'12; Terry⁺ SOSP'13; Balegas⁺ EuroSys'15; Li⁺ USENIX ATC'18
- Stronger operations require synchronisation between replicas
- Pay for stronger semantics with latency, possible unavailability and money

Consistency choices

- Hard to figure out the minimum consistency level necessary to maintain correctness
- Reason about all possible abstract executions?
 - Abstract from some of implementation details, but still describe behaviour of the whole system
 - Number of possible executions is exponential: e.g., choices of vis = order of message deliveries
- Need verification techniques that limit the exponential blow-up

Verification problem

Given

- a set of operations: withdraw(), deposit(), ...
- a conflict relation: *withdraw* 🛛 *withdraw*

Do the operations always preserve a given integrity invariant?

 $I = (balance \geq 0)$

Verification problem

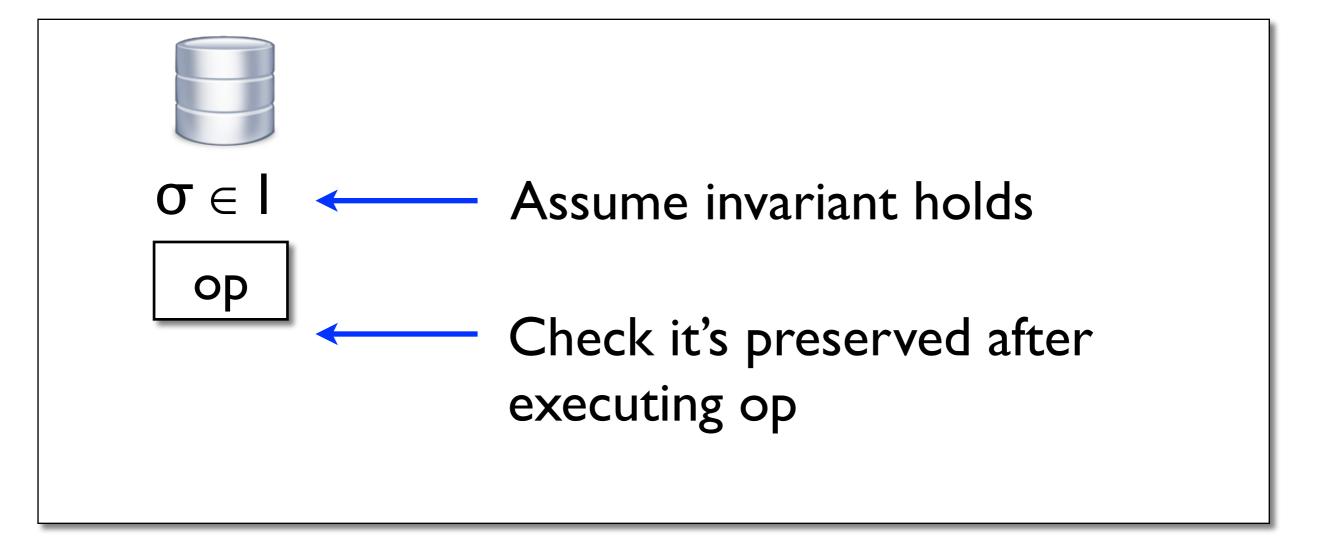
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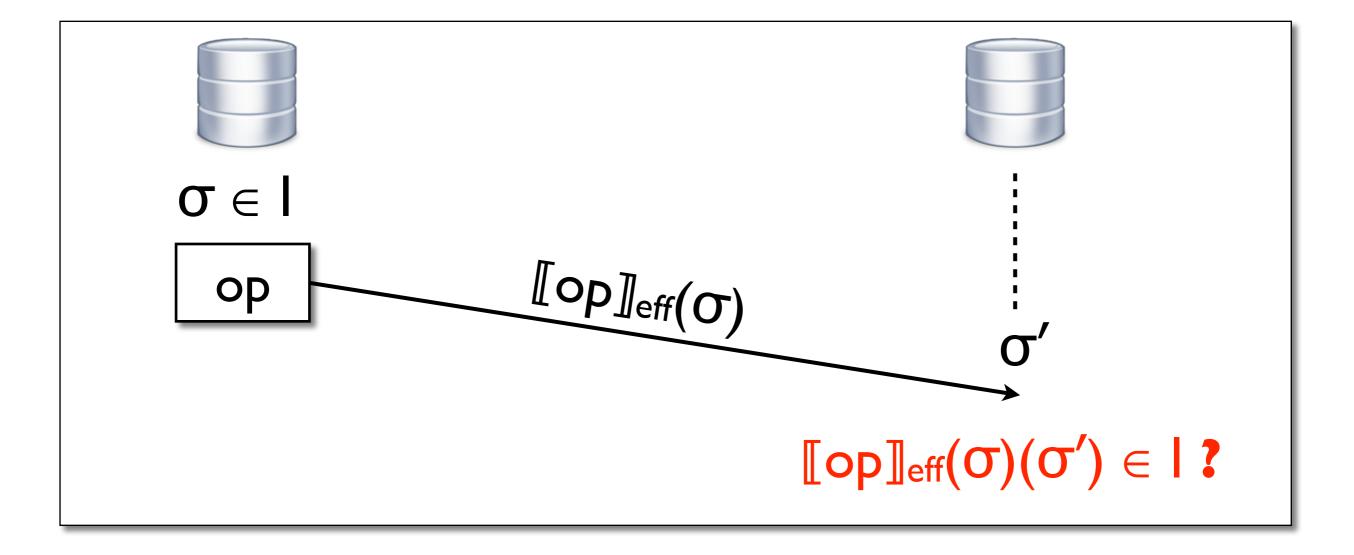
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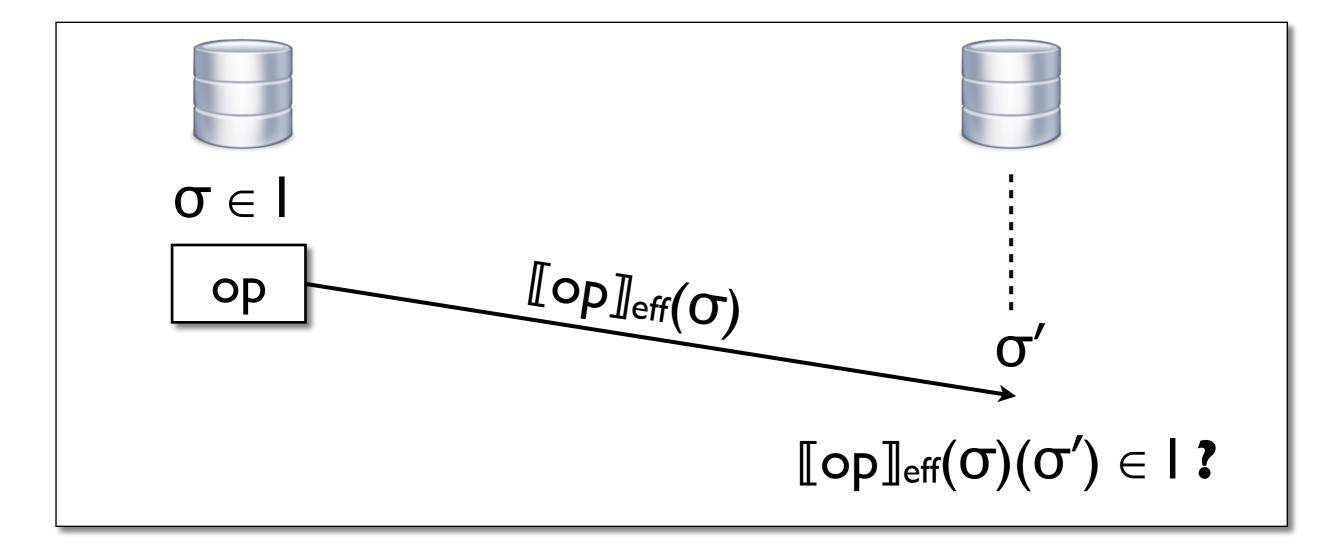
Later: operations -> whole transactions



Single check: no state-space explosion from concurrency

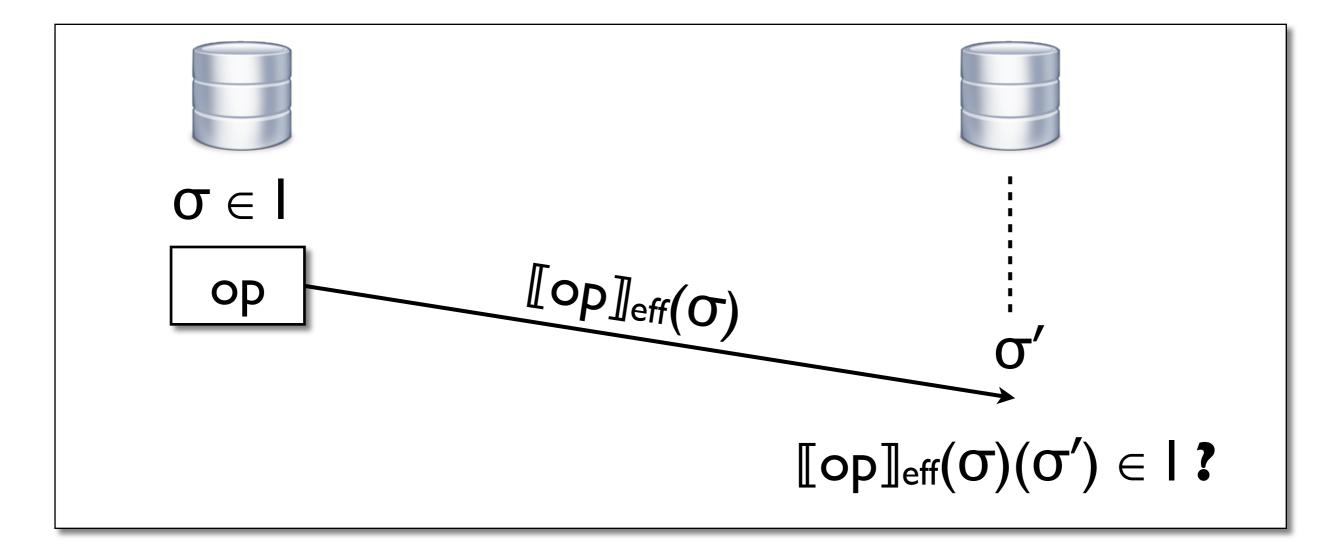


Effect applied in a different state!

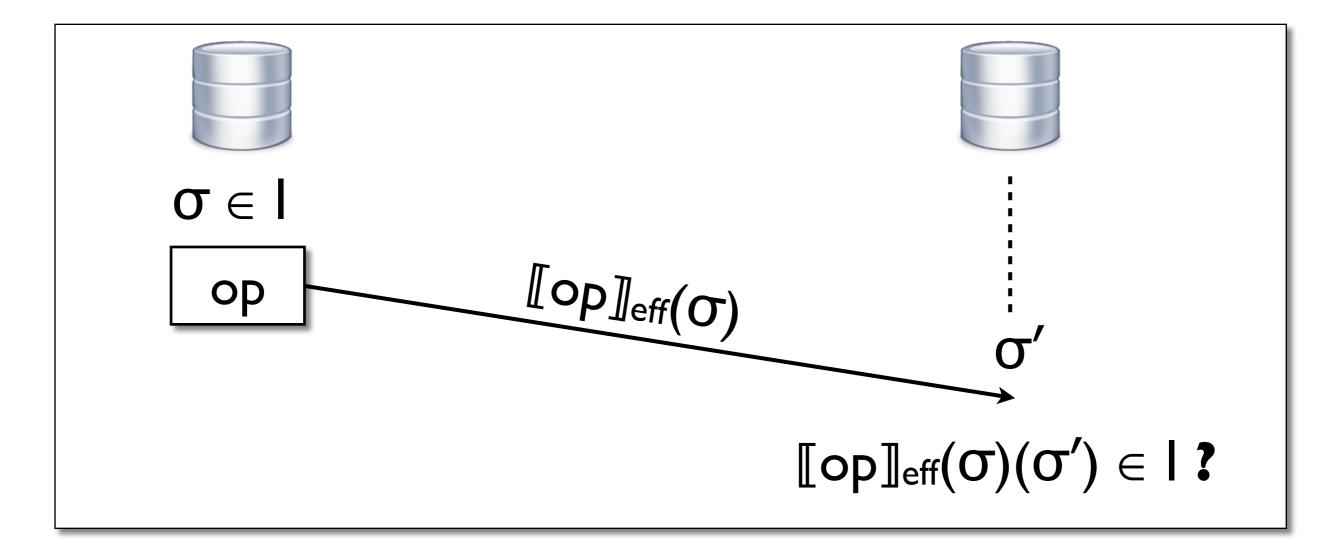


 $[op]_{eff}(\sigma) = if P(\sigma) then f(\sigma) else if...$

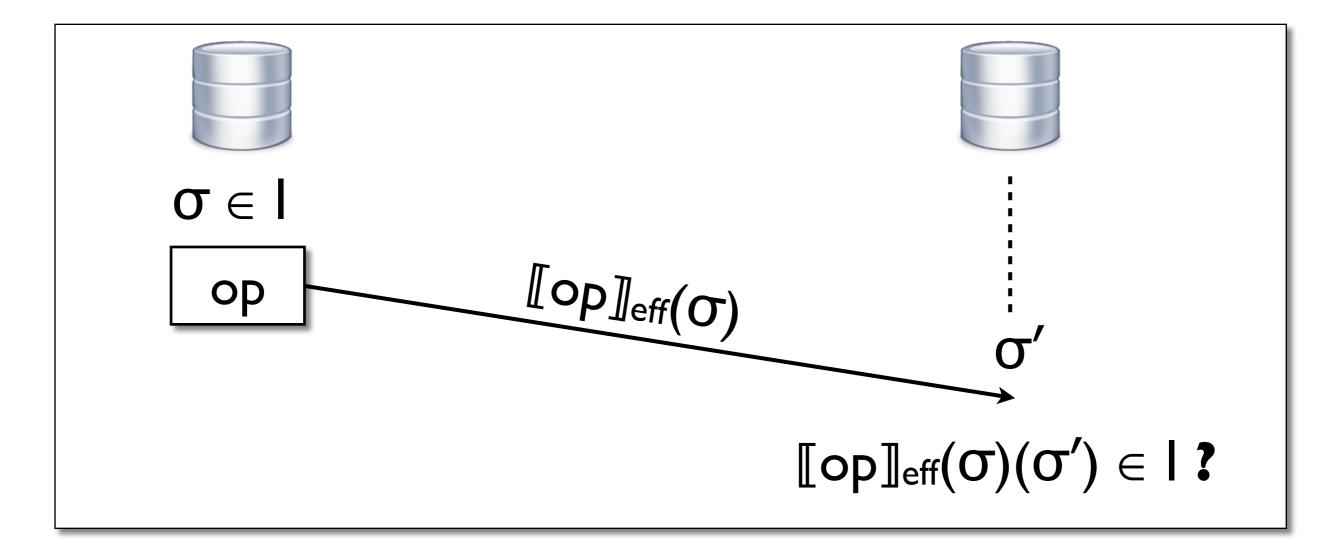
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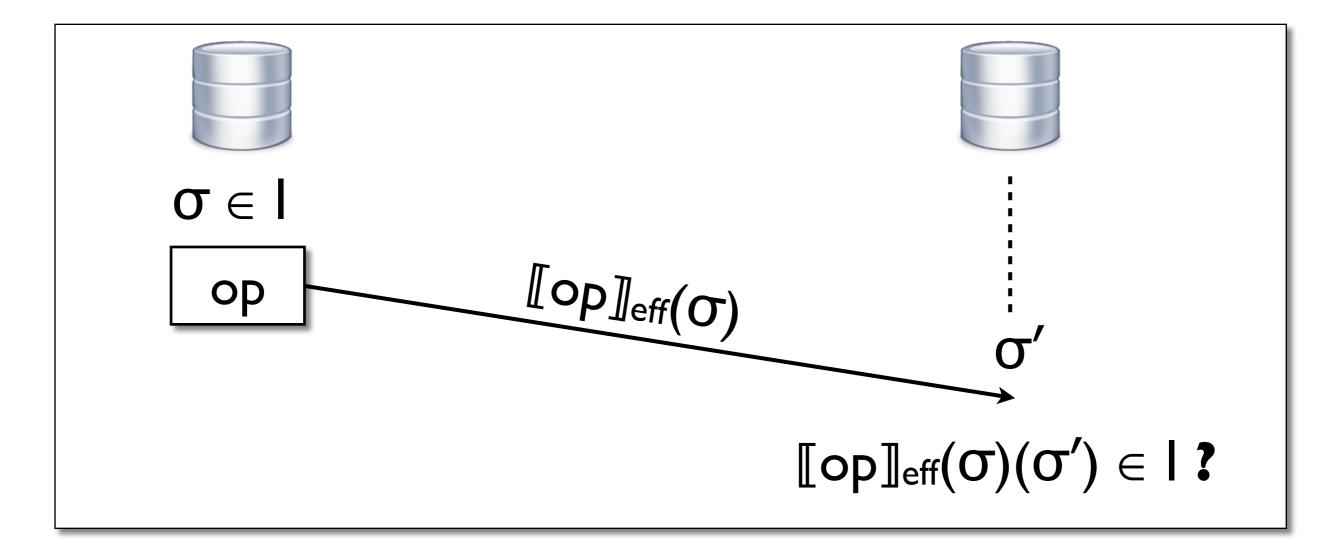
I. Effector safety: $f(\sigma)$ preserves I when executed in any state satisfying P: $\{I \land P\} f(\sigma) \{I\}$



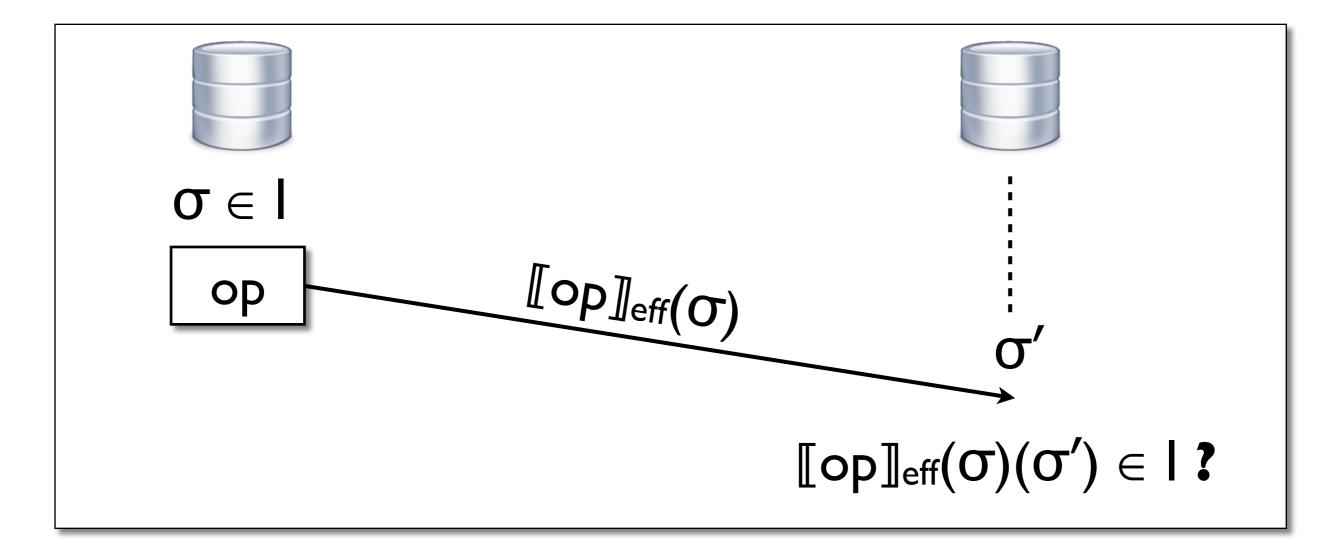
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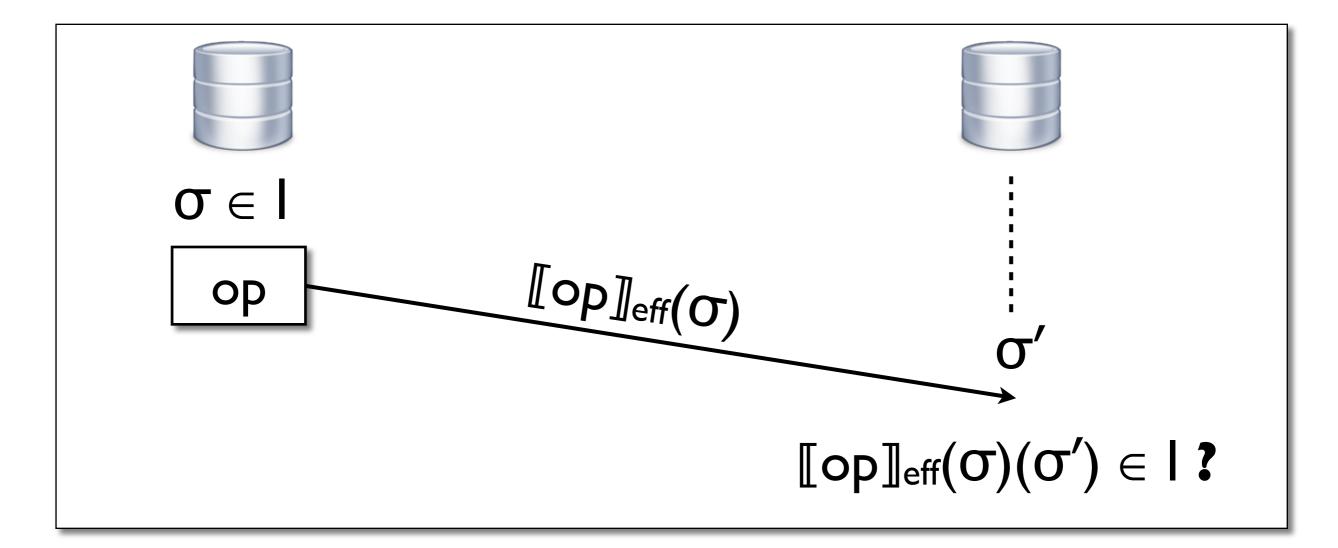
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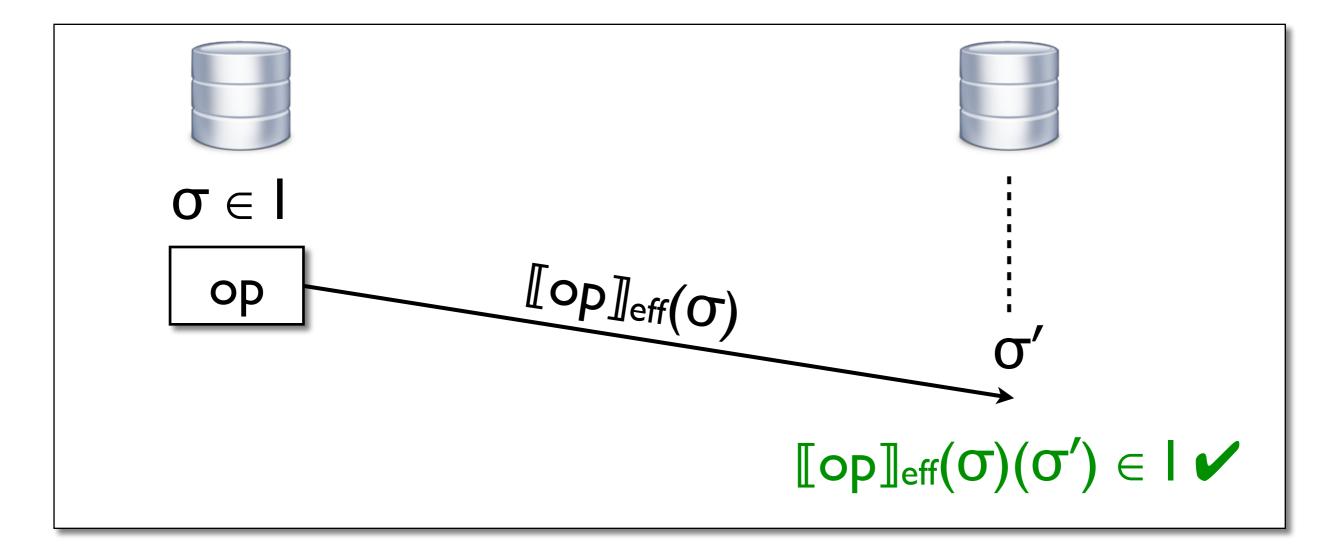
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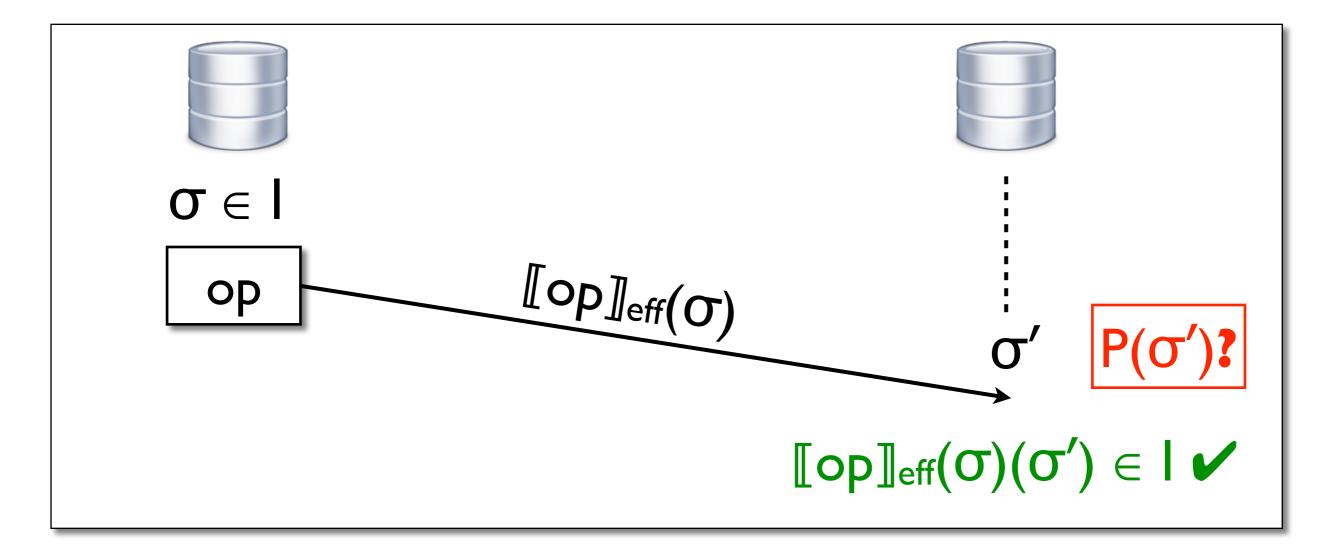
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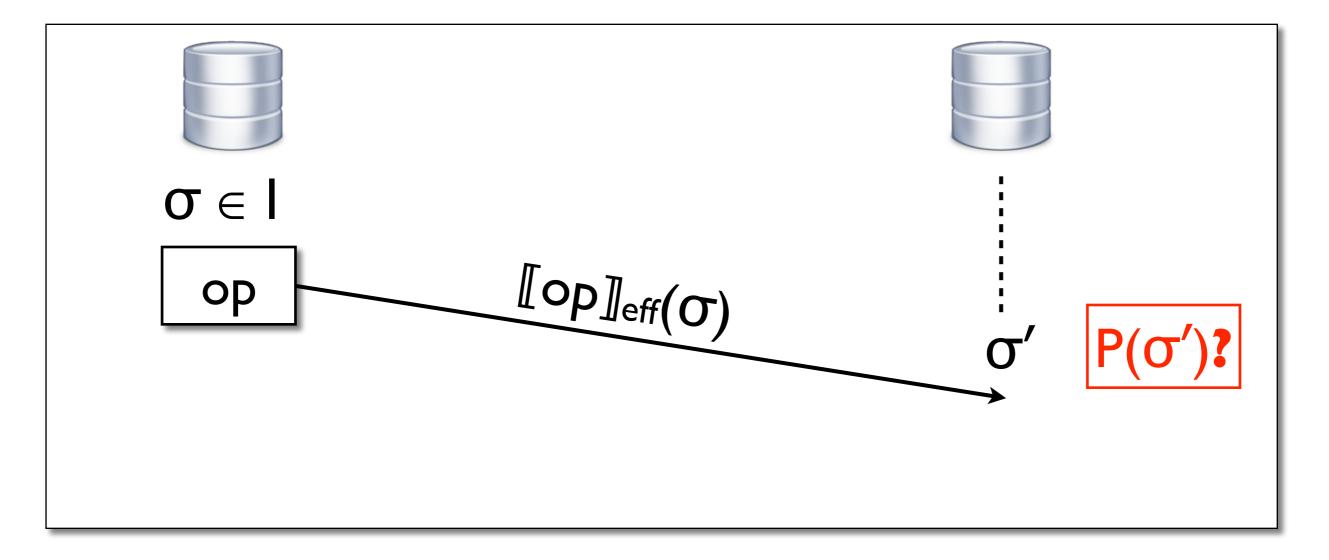
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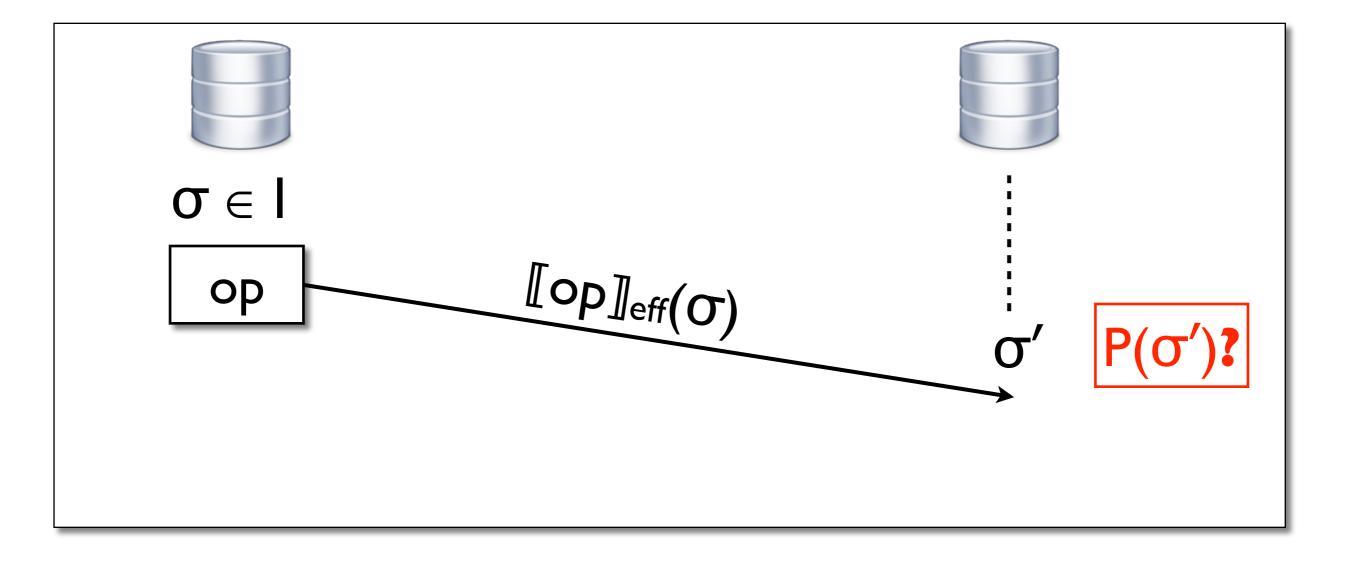
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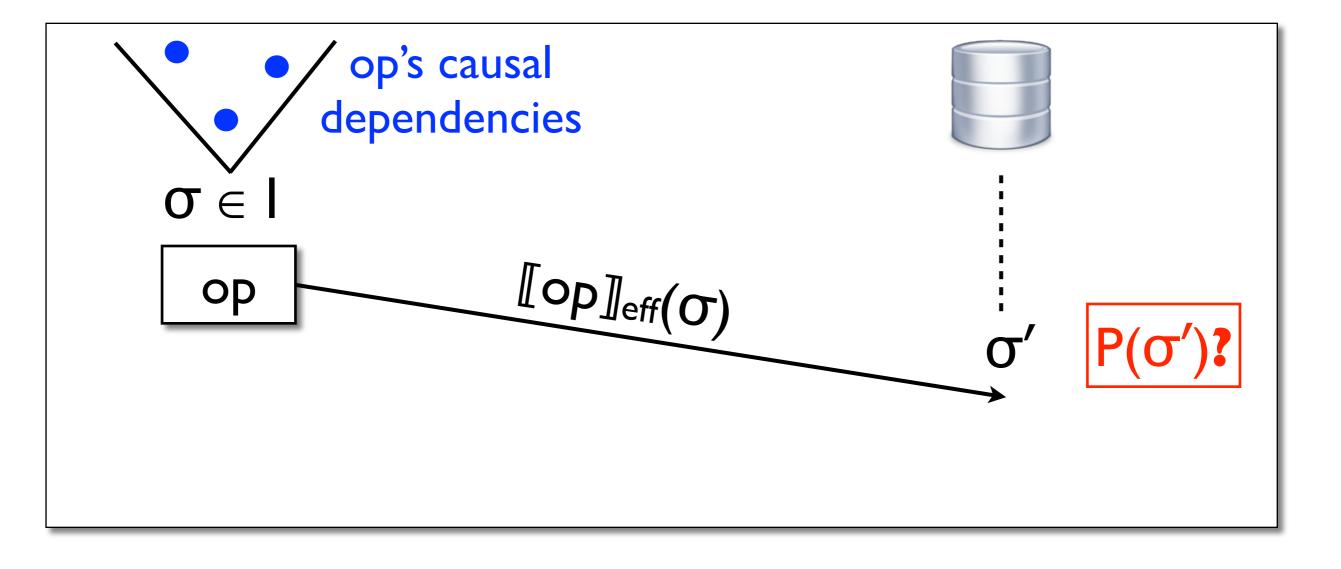


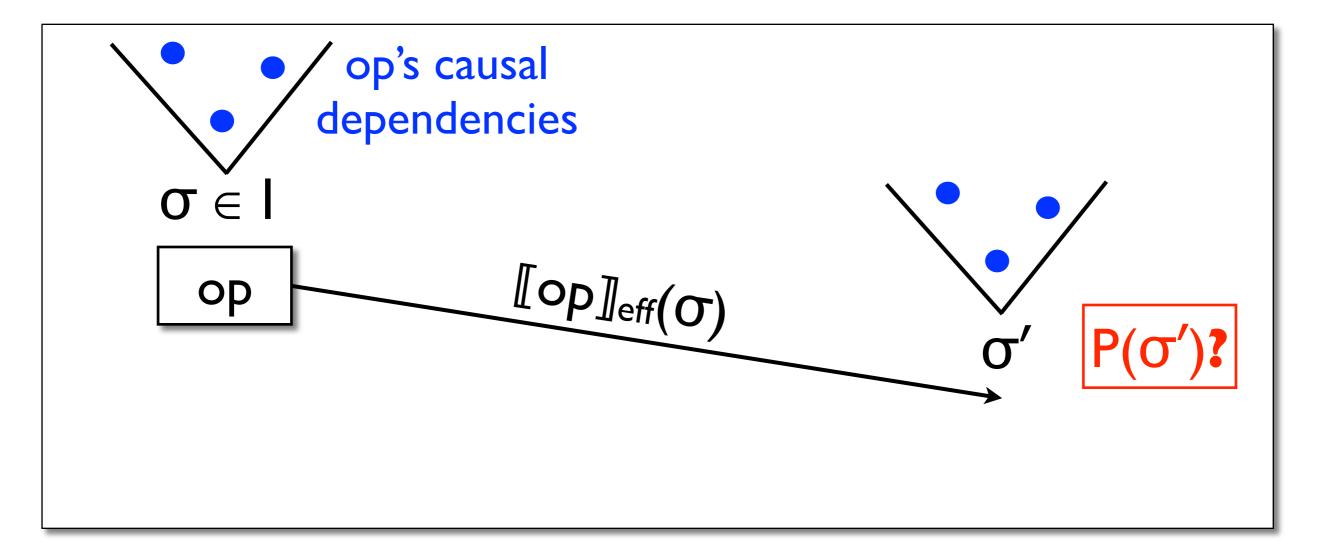
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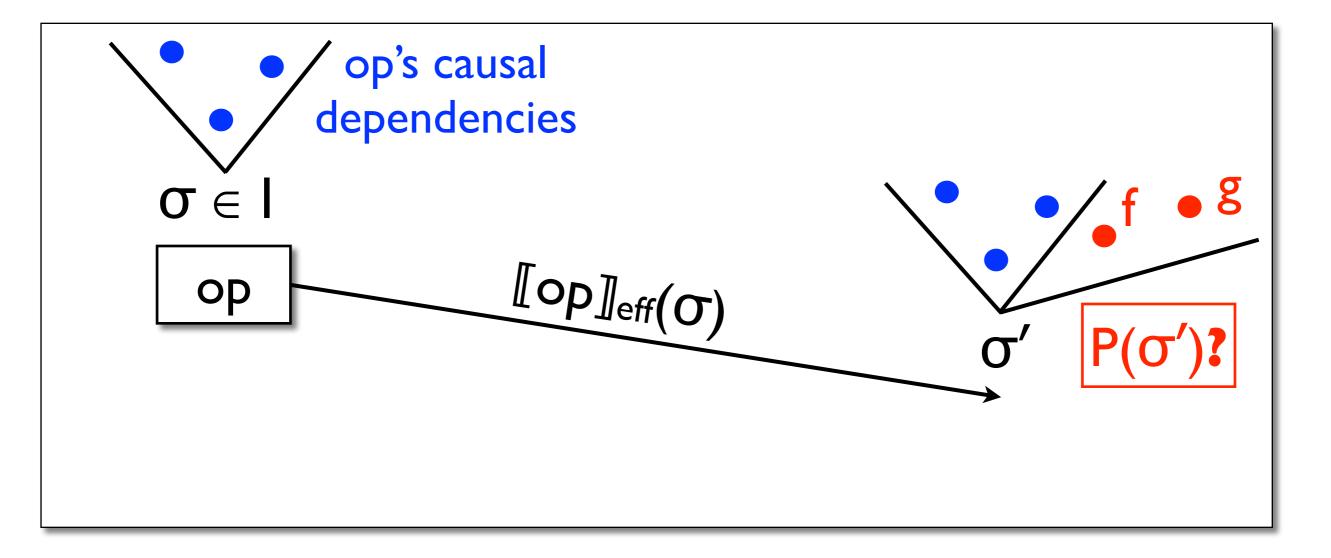


- I. Effector safety: $f(\sigma)$ preserves I when executed in any state satisfying P: $\{I \land P\} f(\sigma) \{I\}$
- 2. Precondition stability: P will hold when $f(\sigma)$ is applied at any replica

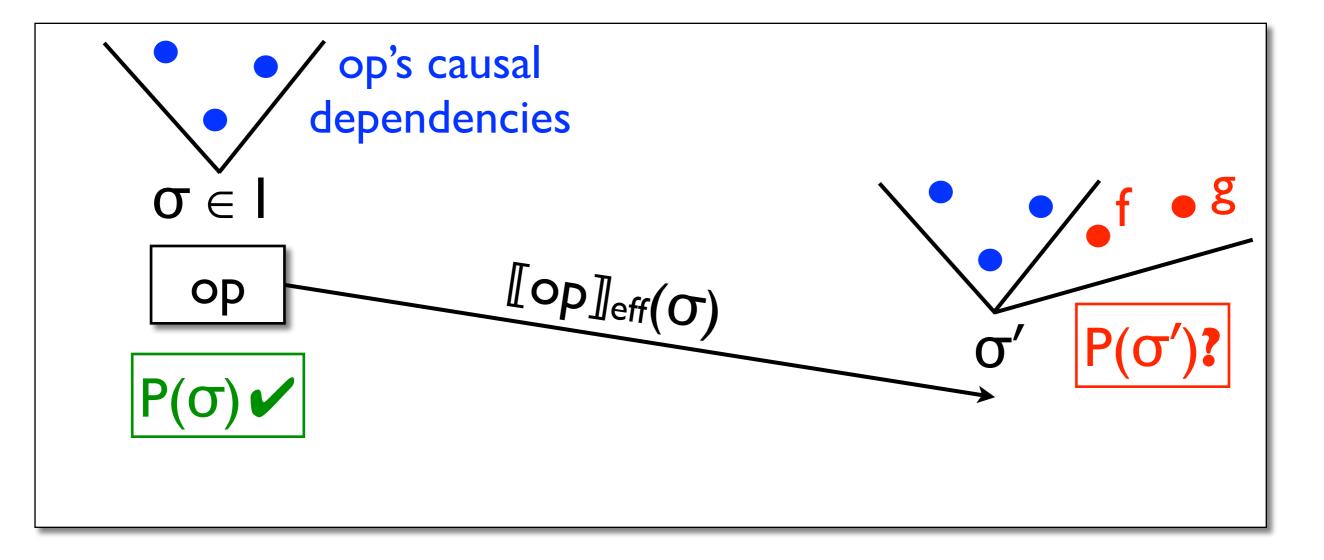








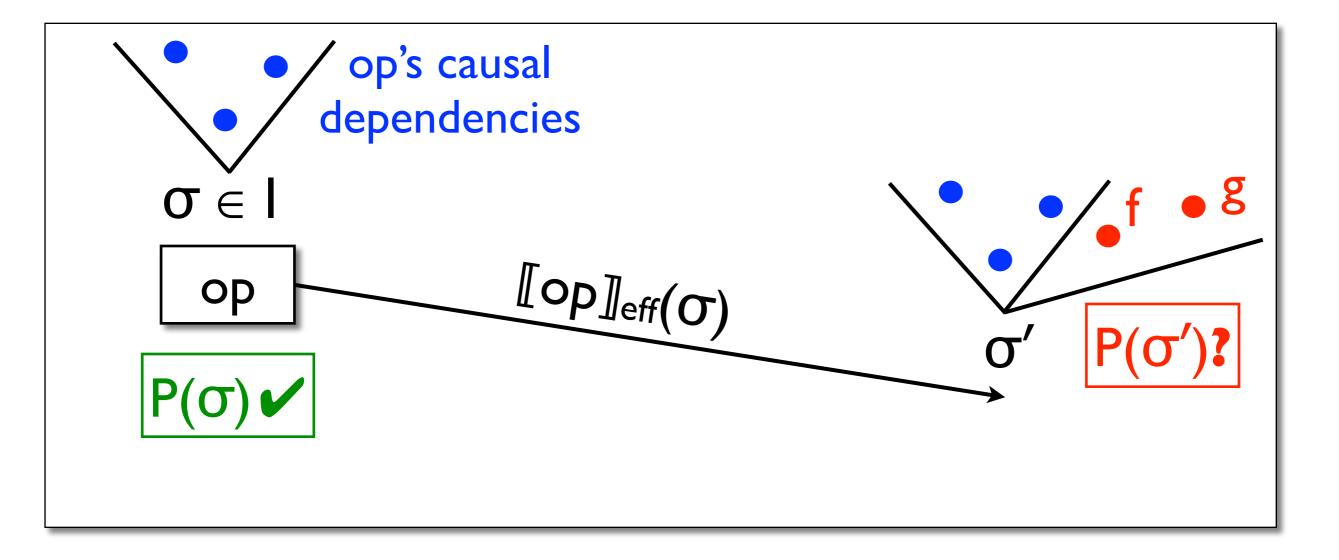
- But can have additional effectors of operations concurrent with op: f, g, ...
- Effectors commute, so $\sigma' = (f; g; ...)(\sigma)$

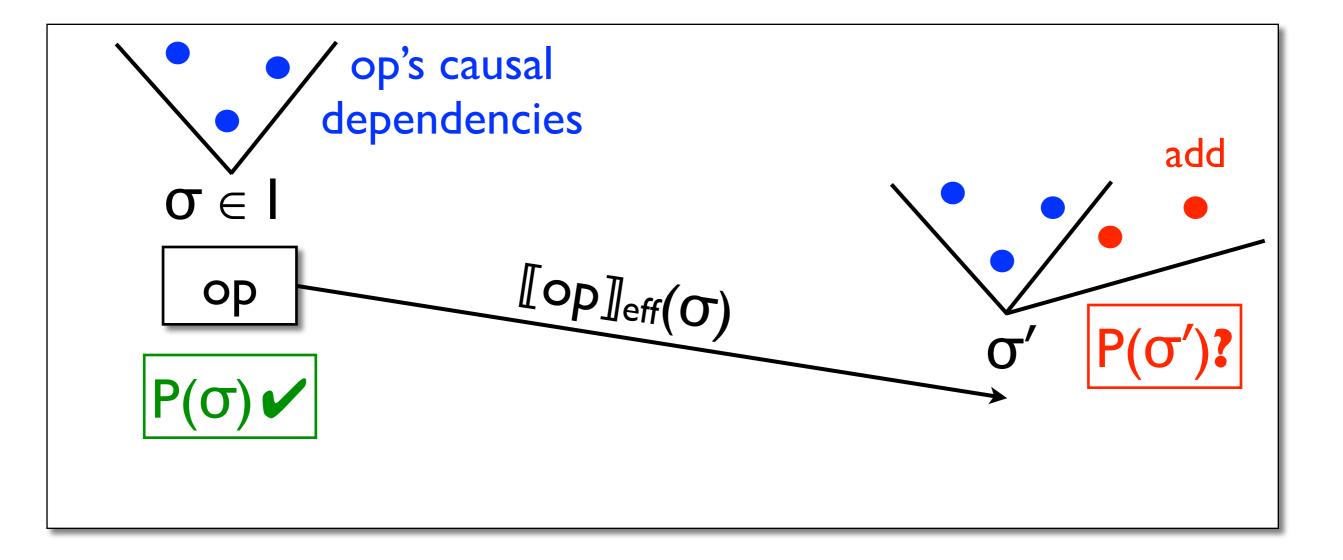


- Causal consistency

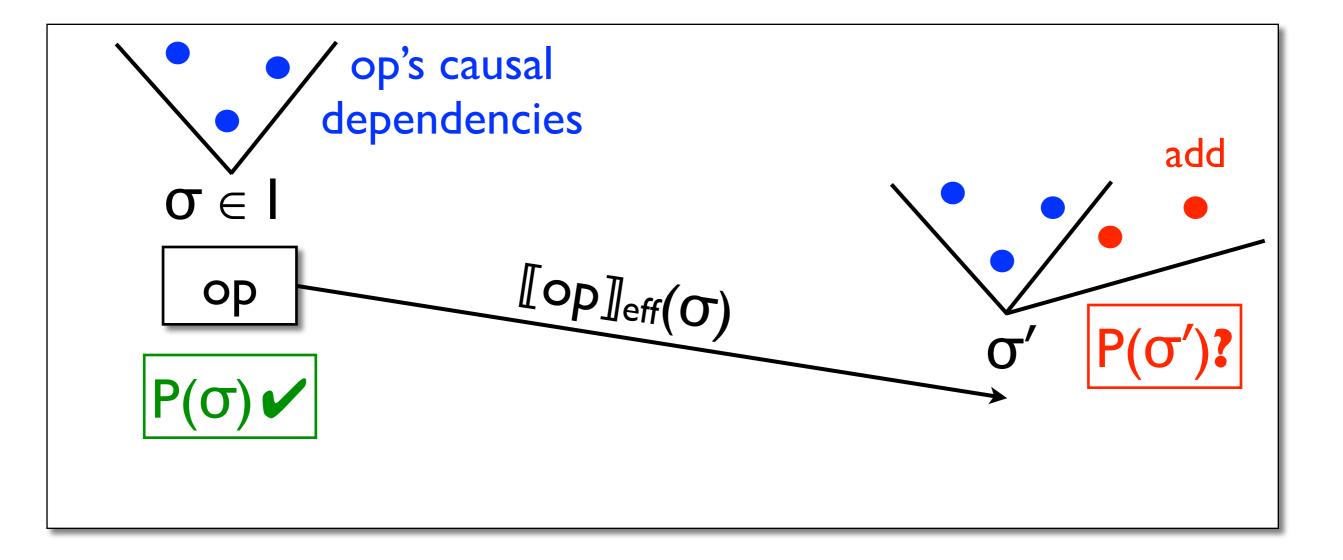
 receive op's causal

 dependencies before receiving op
- But can have additional effectors of operations concurrent with op: f, g, ...
- Effectors commute, so $\sigma' = (f; g; ...)(\sigma)$

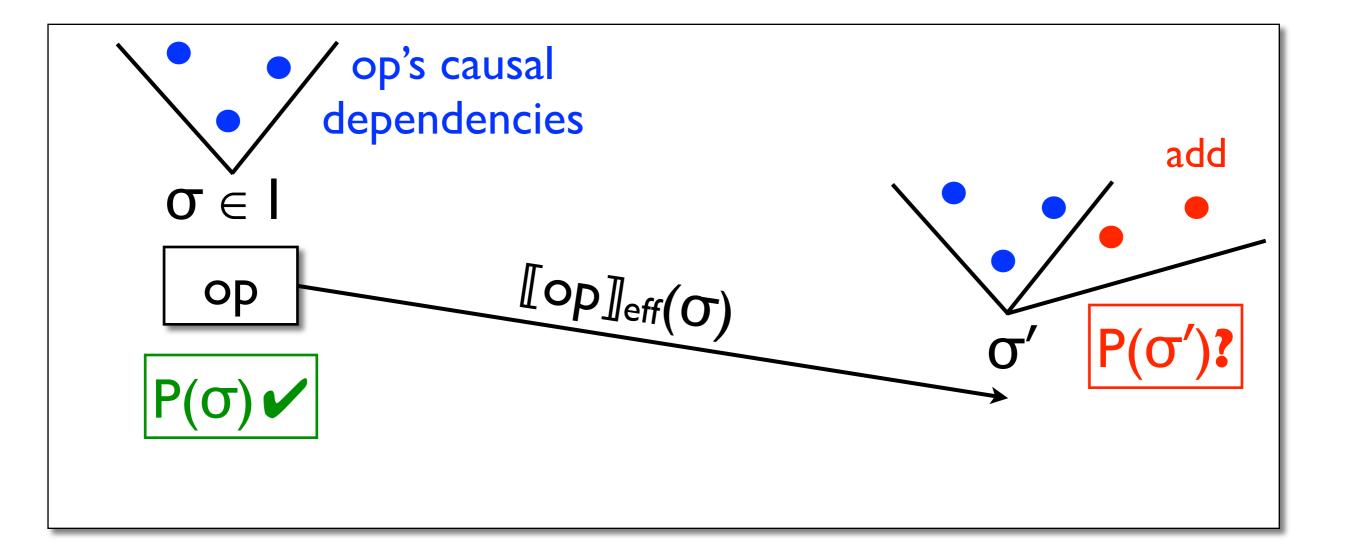




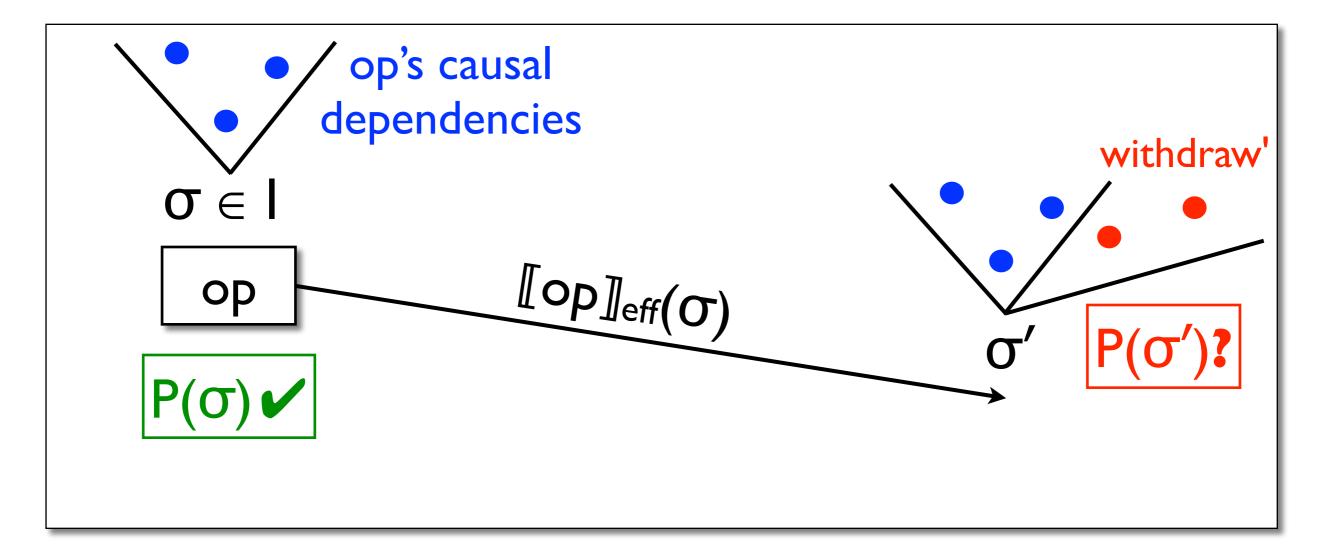
 $\{bal \ge 100\} \ bal := bal + 100 \ \{bal \ge 100\}$



 $\{bal \ge 100\}$ bal := bal+100 $\{bal \ge 100\}$

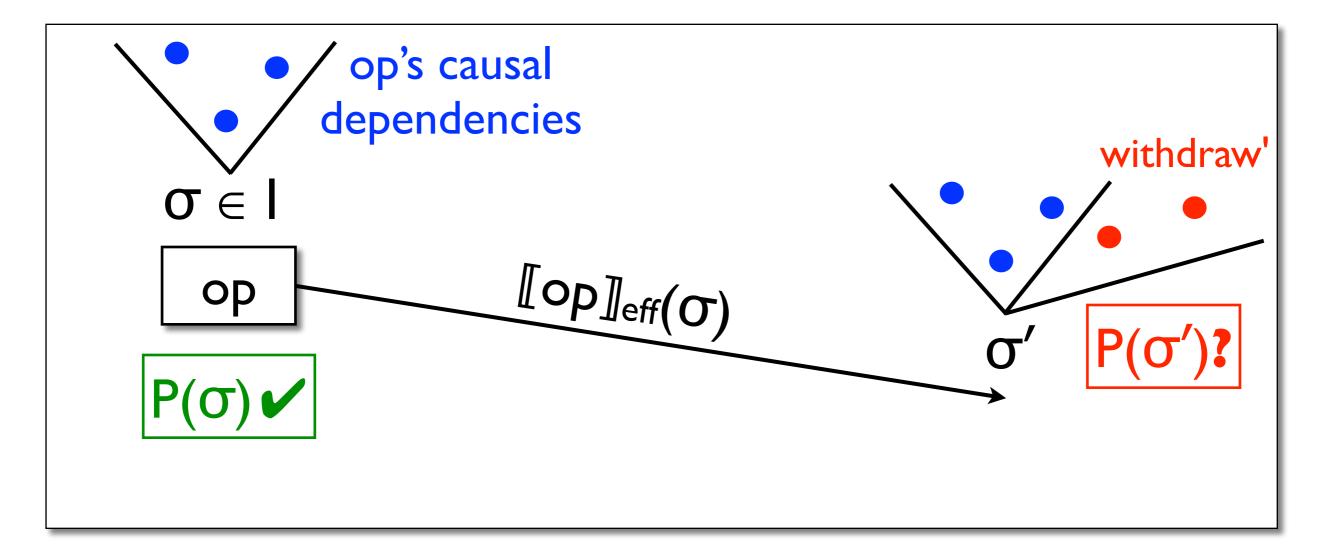


 $\{bal \ge 100\} bal := bal + 100 \{bal \ge 100\}$



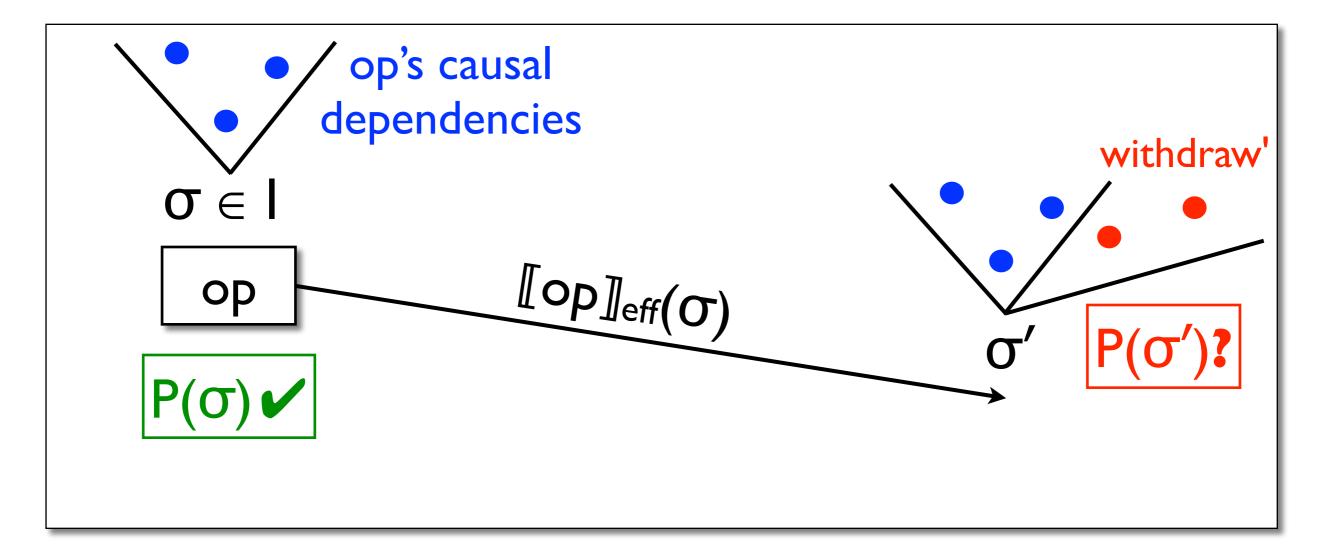
 $\{bal \ge 100\} bal := bal+100 \{bal \ge 100\}$

 $\{bal \ge 100\} \ bal := bal - 100 \ \{bal \ge 100\}$



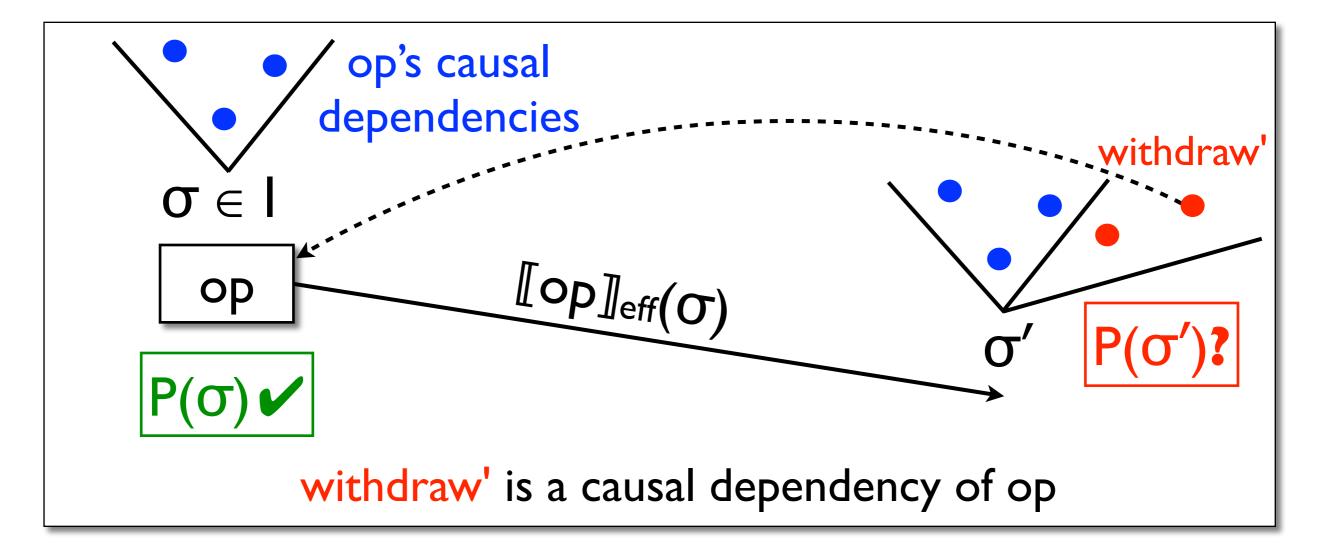
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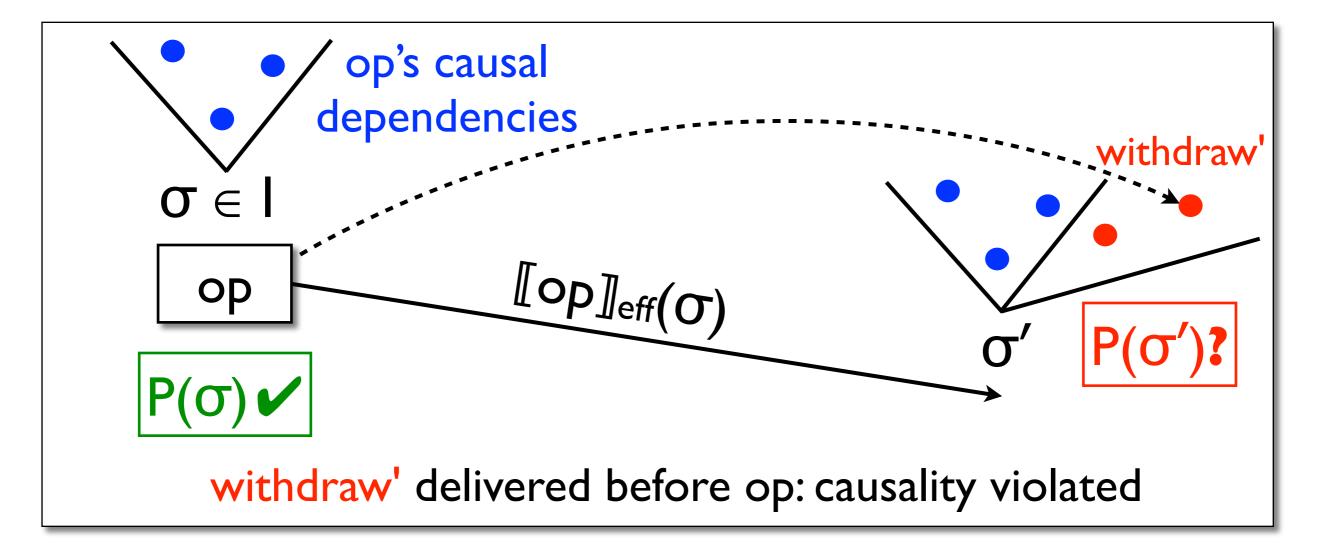


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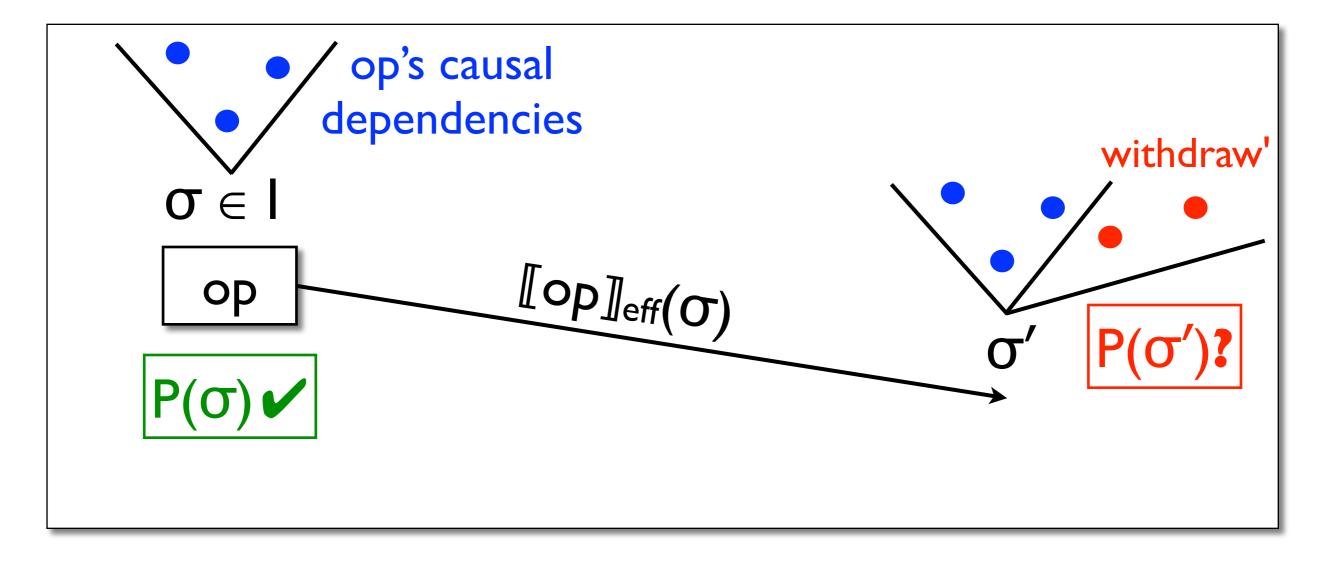
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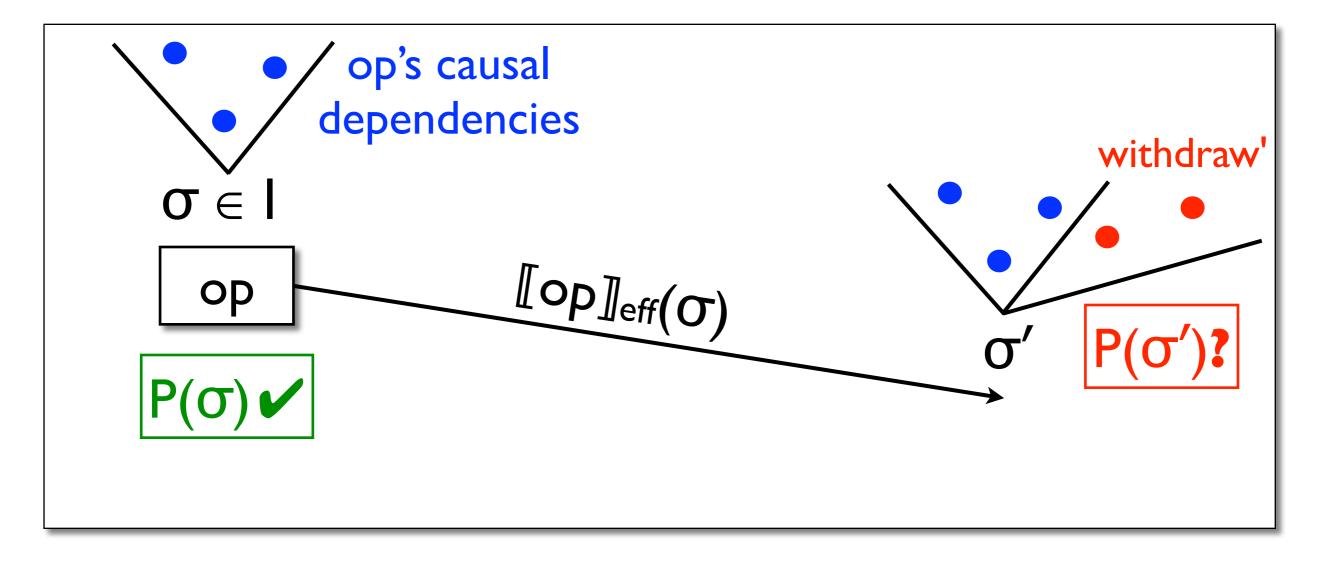
withdraw \bowtie withdraw; $\neg(add \bowtie withdraw)$



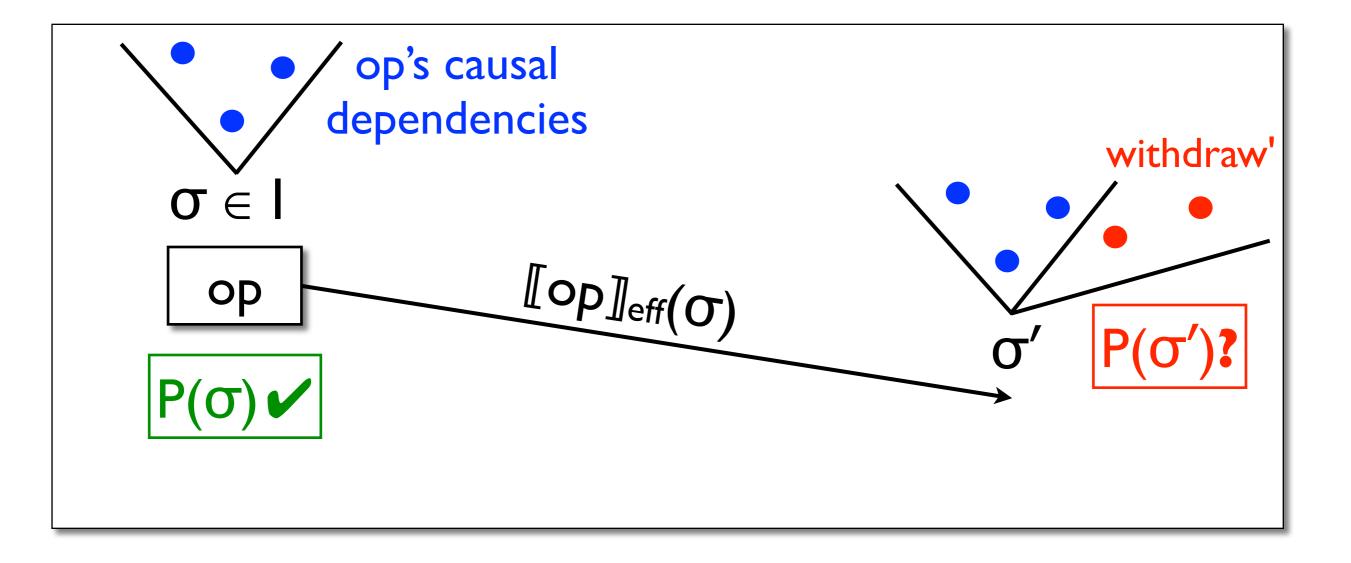
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withdraw \bowtie withdraw; $\neg(add \bowtie withdraw)$



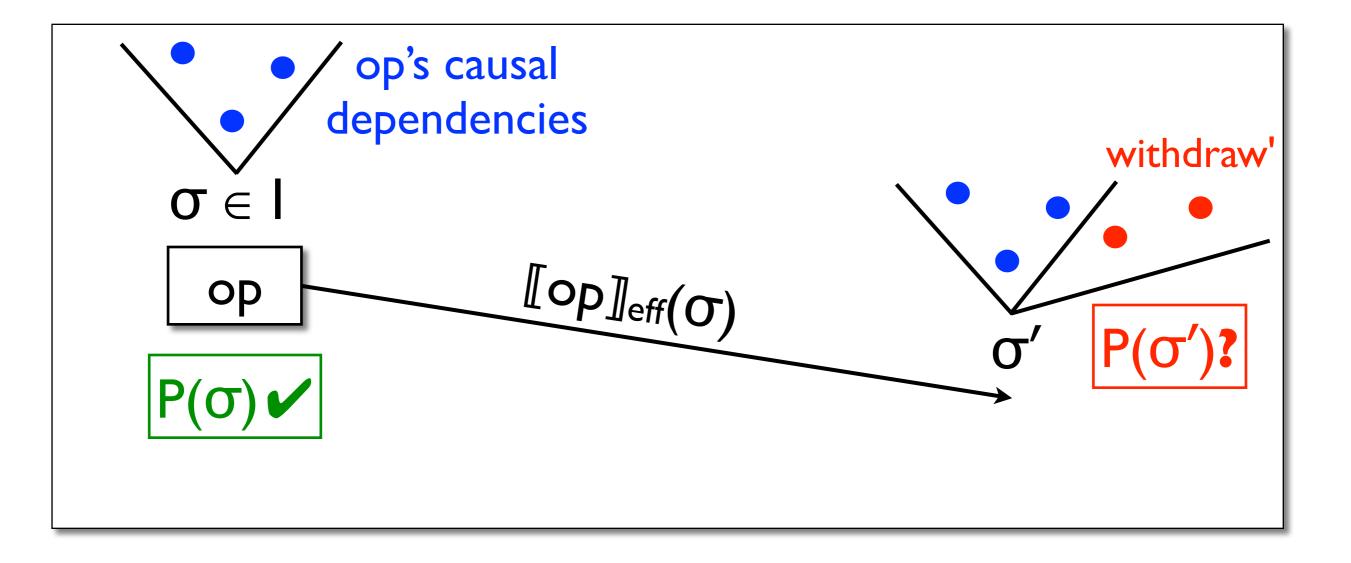
Only requires checking each pair of operations: no exponential explosion!



Can infer the conflict relation \bowtie : op₁ \bowtie op₂ if the precondition of op₁ unstable under the effector of op₂

Pre of withdraw under effector of add:

 $\{bal \ge 100\} bal := bal+100 \{bal \ge 100\}$, no \bowtie



Can infer the conflict relation \bowtie : op₁ \bowtie op₂ if the precondition of op₁ unstable under the effector of op₂

Pre of withdraw under effector of withdraw:

 $\{bal \ge 100\}$ bal := bal-100 $\{bal \ge 100\}$ X, need \bowtie

Correct Eventual Consistency Tool

- Developed by Sreeja Nair (UPMC, Paris)
- Model application in a domain-specific language, including replicated data type libraries
- Model compiled into a Boogie program encoding the conditions of the proof rule
- Discharged using SMT
- Automatically infers a conflict relation

https://github.com/LightKone/correct-eventualconsistency-tool

Demo

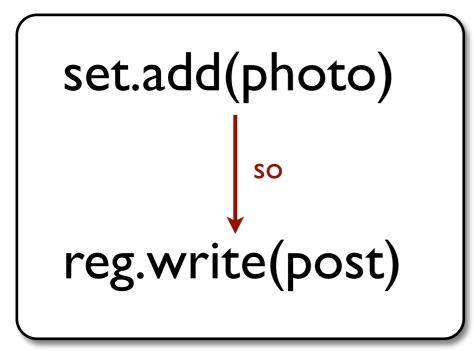
Transactions

Transactions

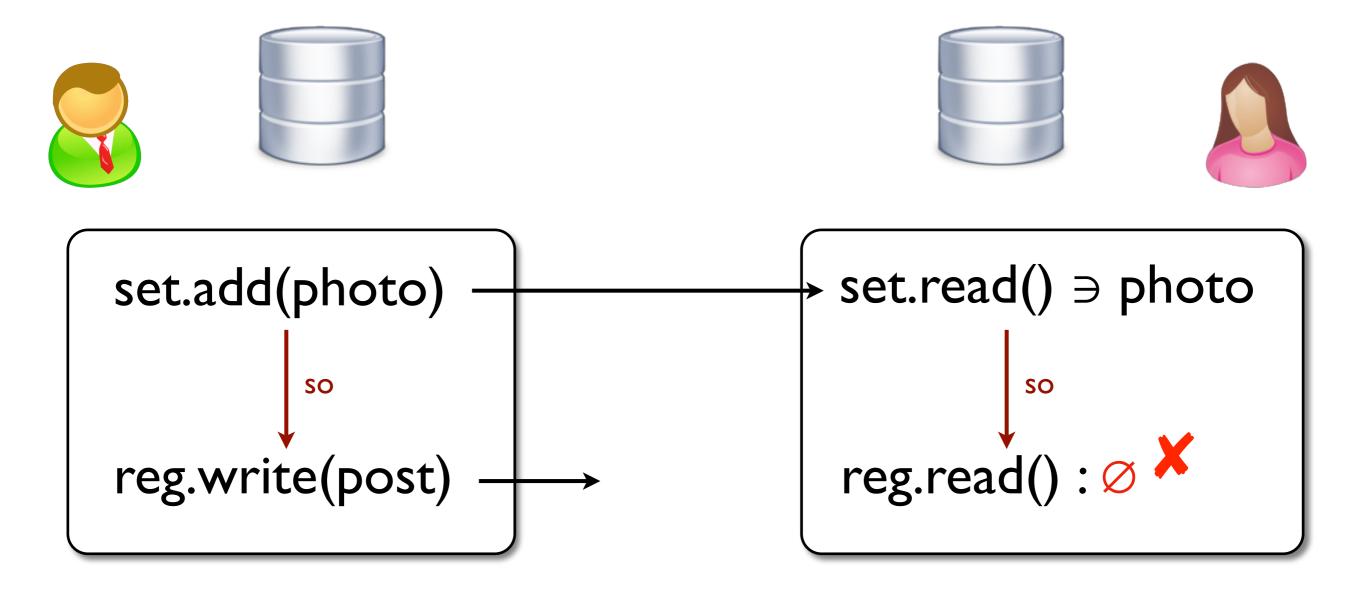
- Fundamental abstraction in databases
- Allow clients to group operations to be processed indivisibly
- Provided by virtually any single-node SQL database
- NoSQL data stores: starting to reappear

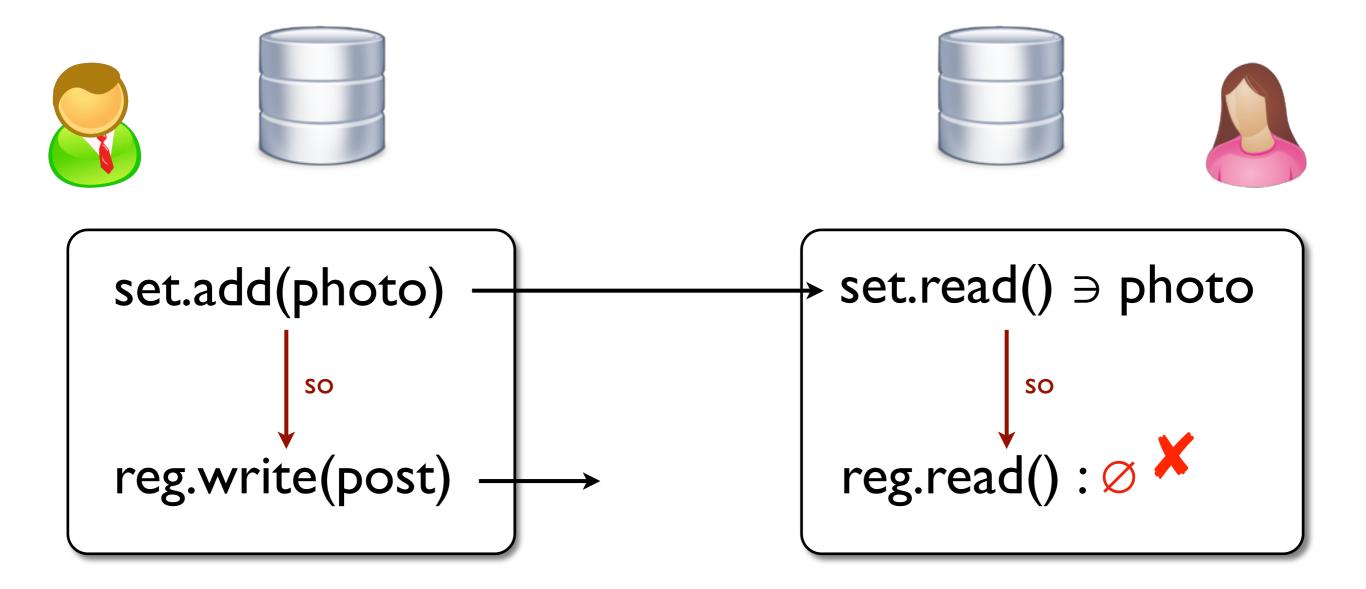




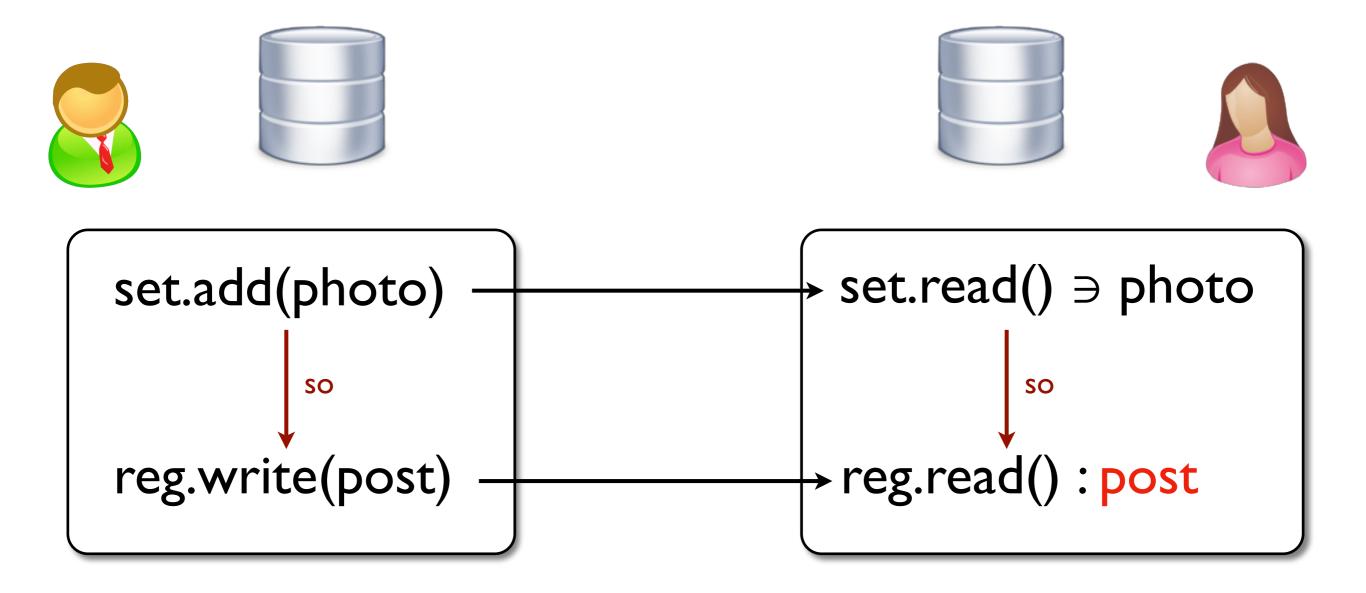


set.read() \ni photo SO reg.read() : \emptyset

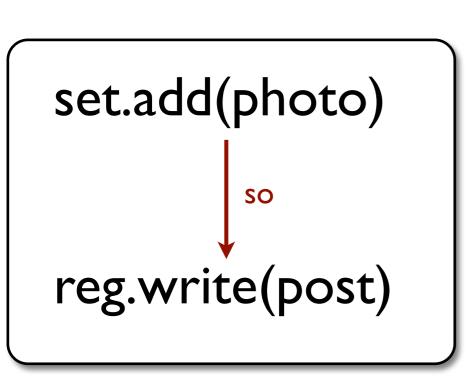




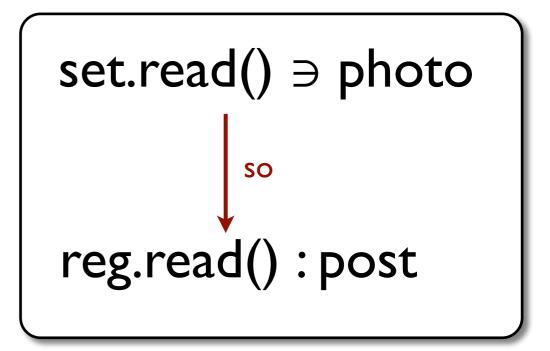
Causal consistency isn't enough

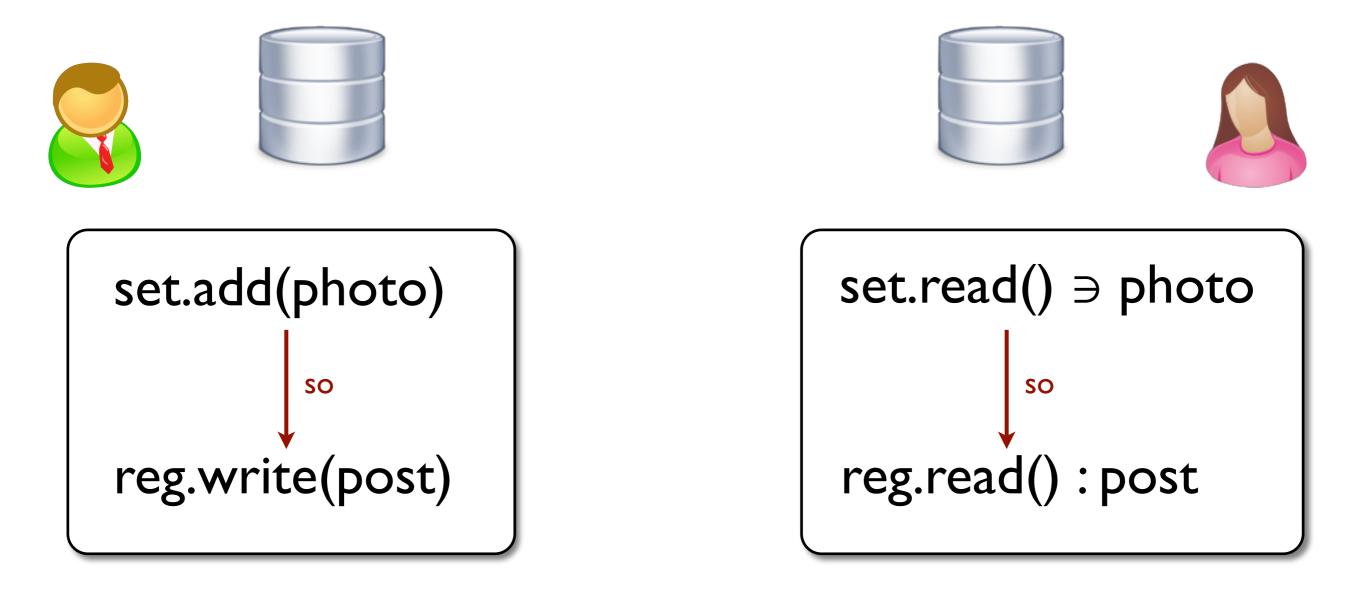




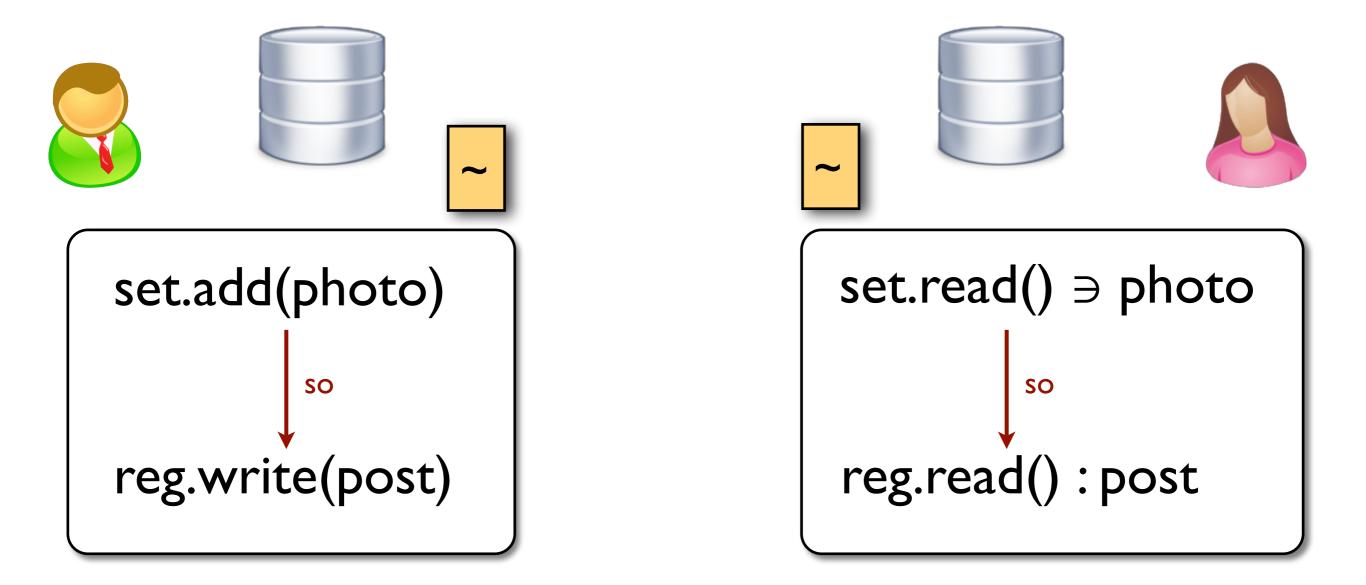




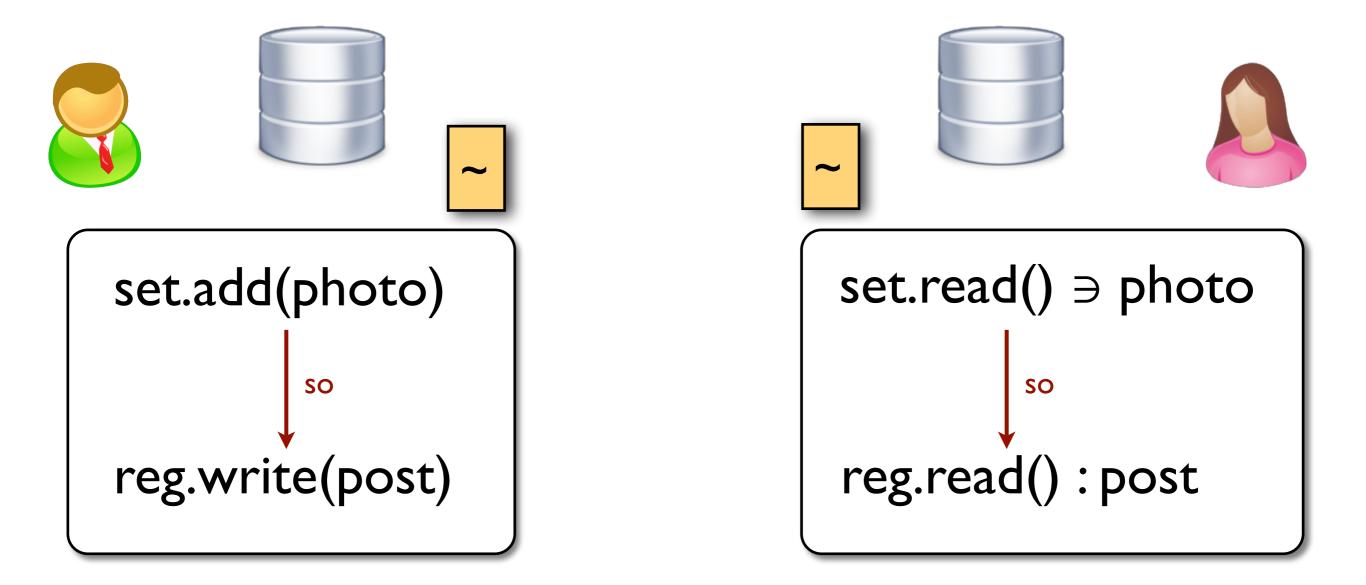




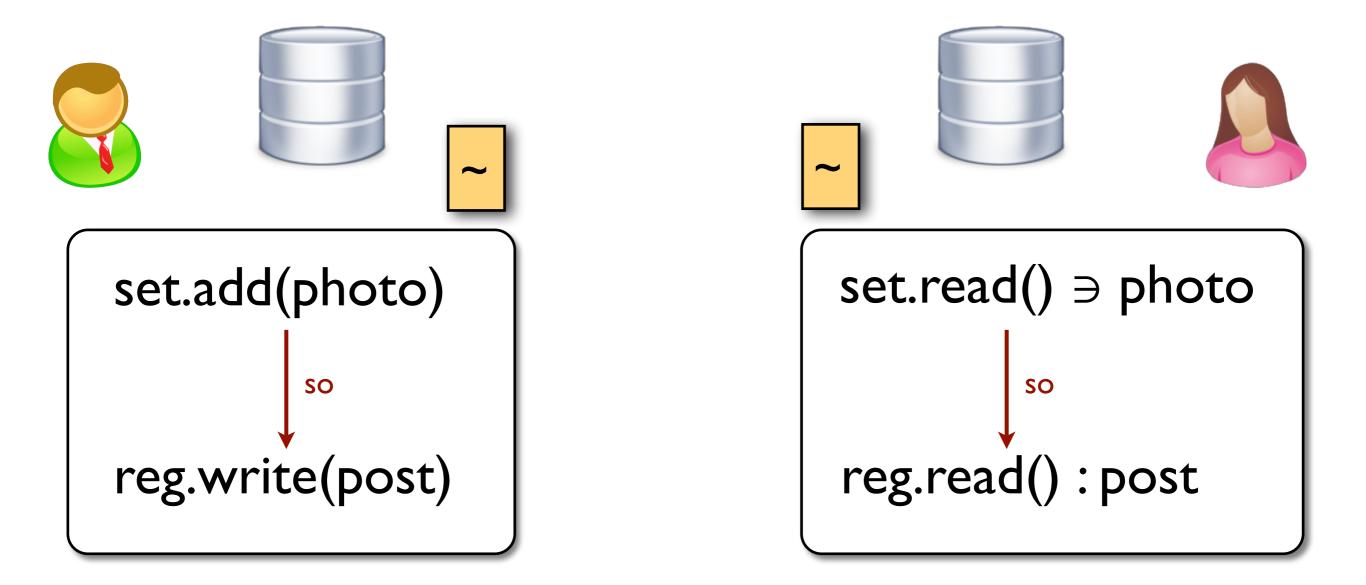
Consistency model = set of histories (E, so, ~)



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- ~: equivalence relation that groups events from the same transaction: transitive, symmetric, reflexive



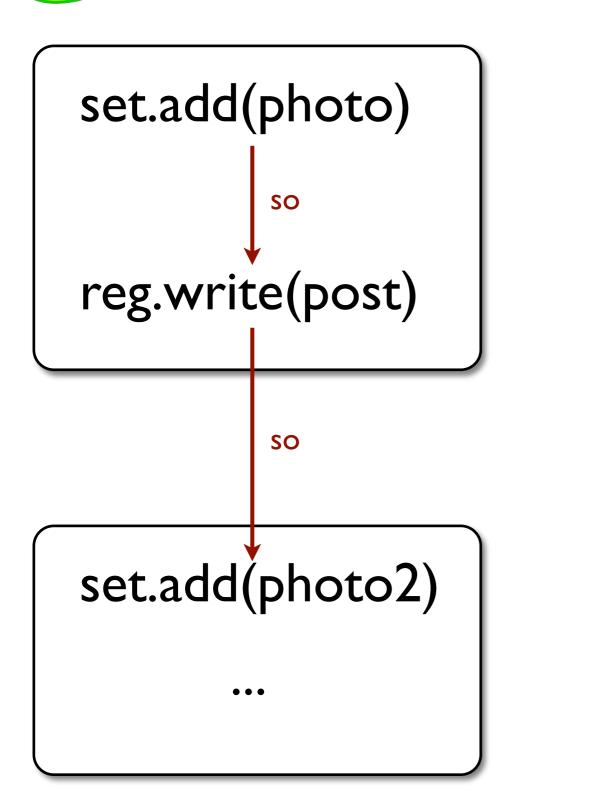
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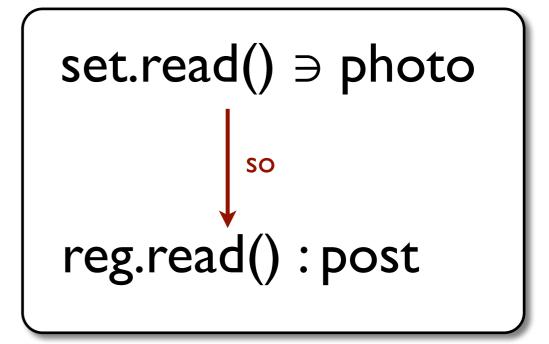


- Consistency model = set of histories (E, so, ~)
- ~: equivalence relation that groups events from the same transaction: transitive, symmetric, reflexive
- For simplicity, assume every transaction completes
- Transaction T: equivalence class of events of ~

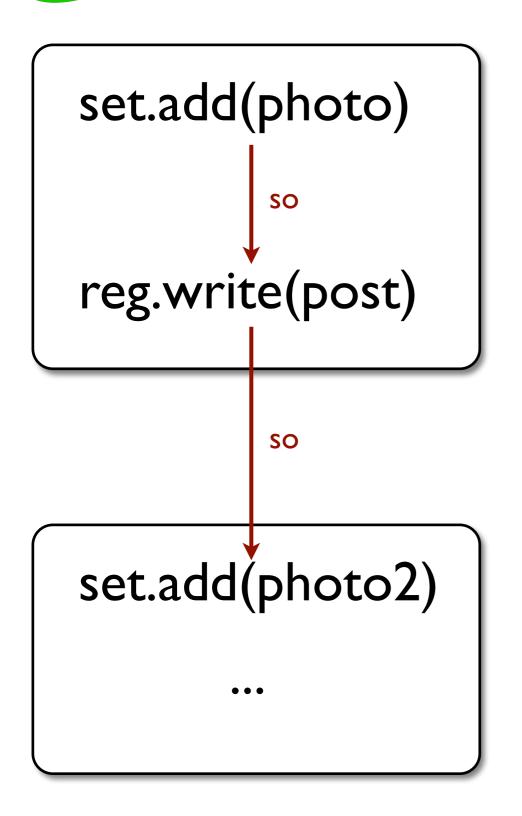


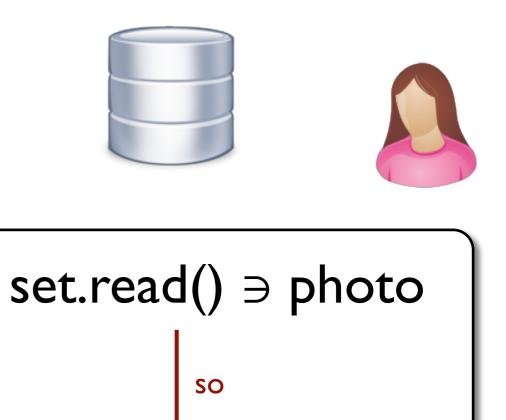












A session is a sequence of transactions: events from the same transaction contiguous in so

reg.read() : post

$$\forall e, f, g \in E. e \xrightarrow{so} f \xrightarrow{so} g \land e \thicksim g$$
$$\implies e \thicksim f \thicksim g$$

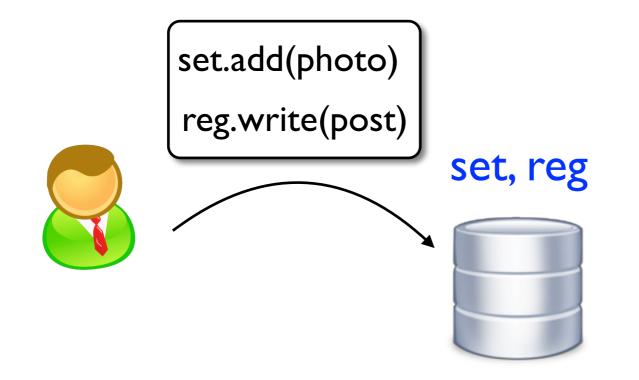
Strongly consistent transactions

Sequential consistency ~ serializability

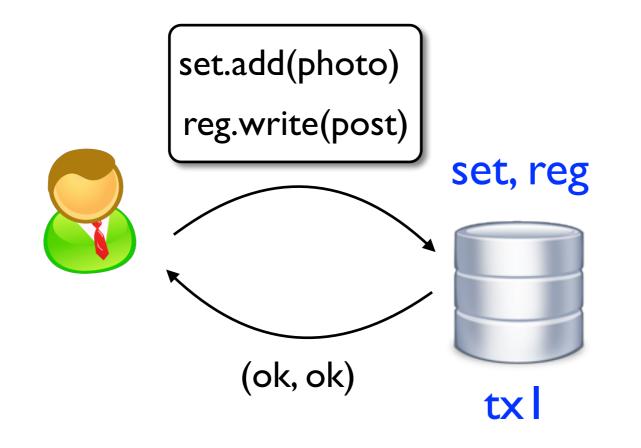




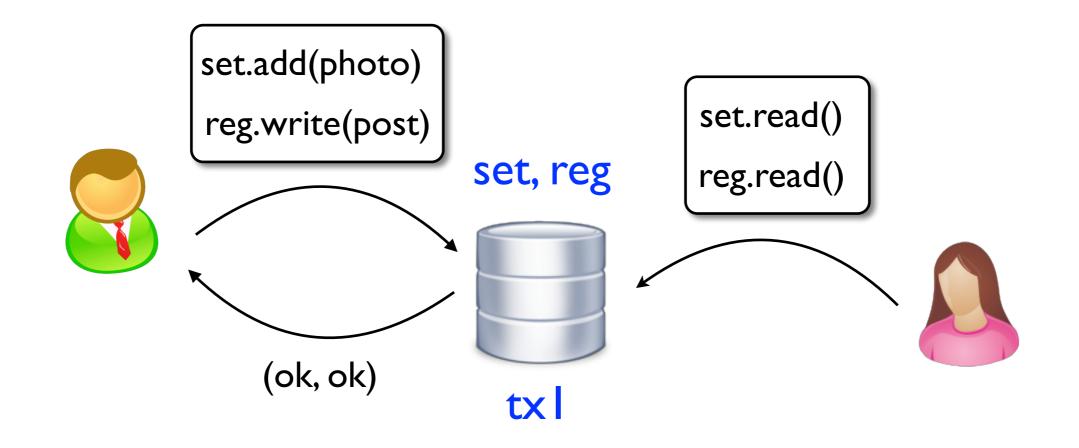
- Server with a single copy of all objects
- Clients send txs to the server and wait for a reply
- Server processes txs atomically in the receipt order



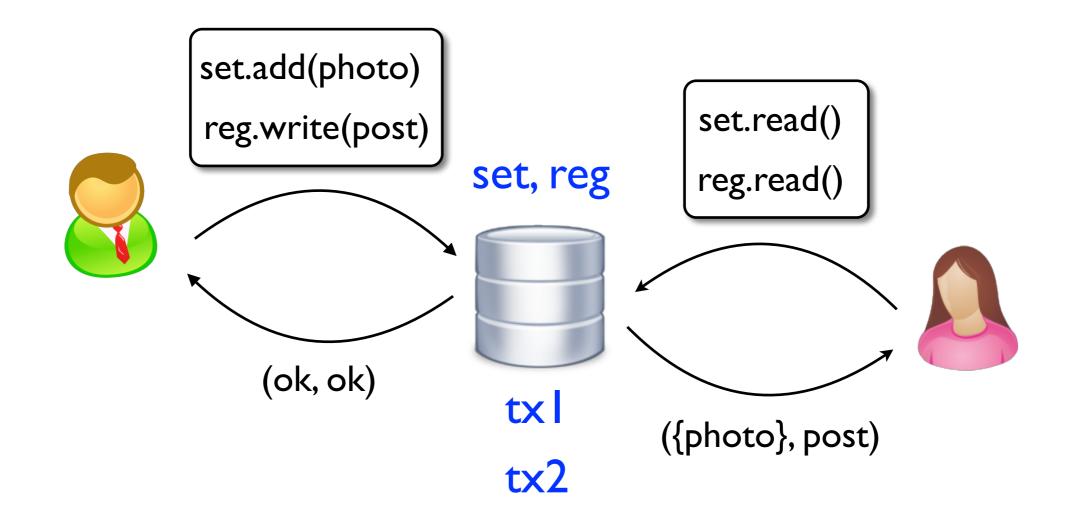
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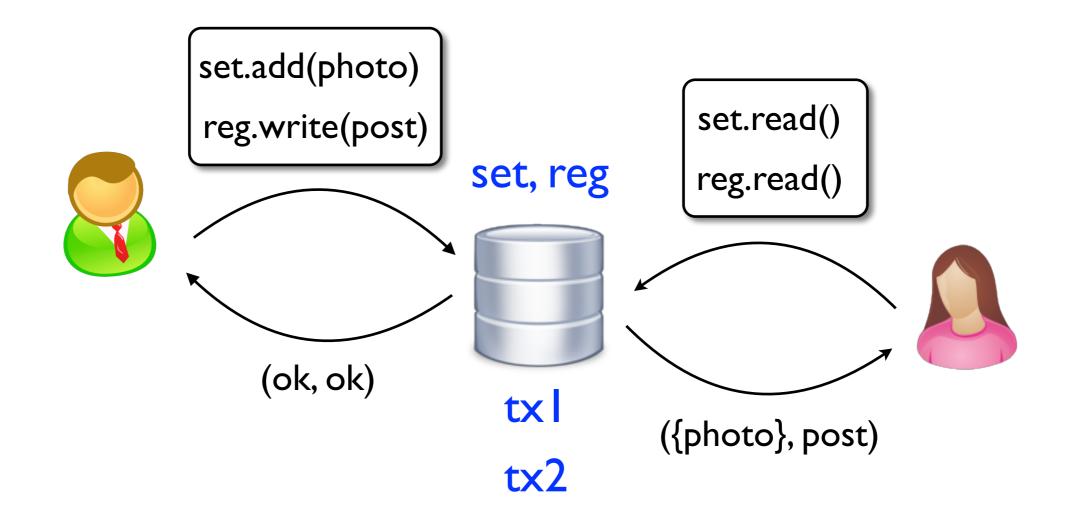
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Serializability = $\{H \mid \exists execution with history H produced by the abstract implementation\}$

Sequential consistency

(E,so) $| \exists$ total order to. (E, so, to) satisfies:

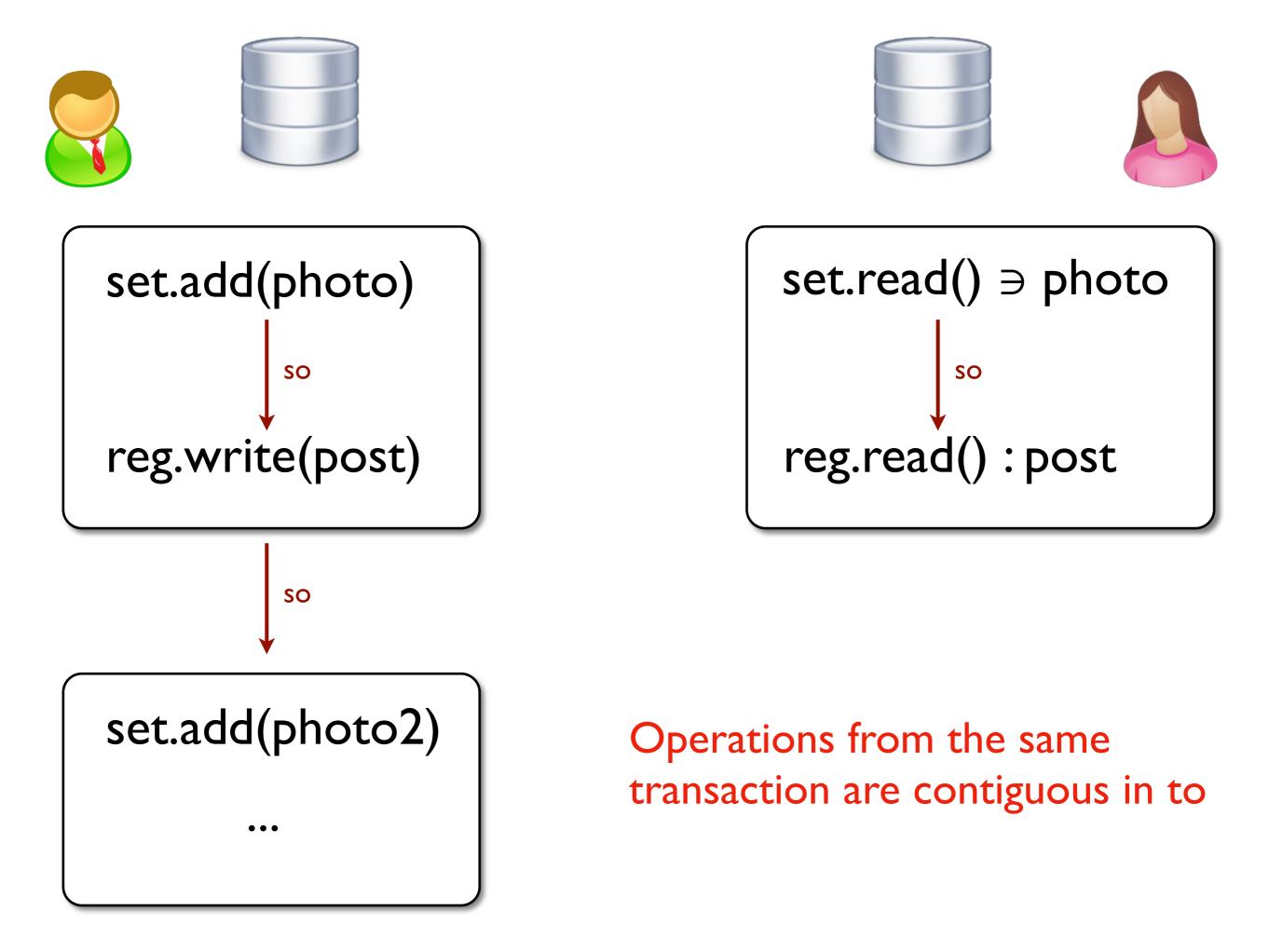
- I. so \subseteq to
- 2. The return value of each operation in E is computed from a state obtained by executing all operations on the same object preceding it in to

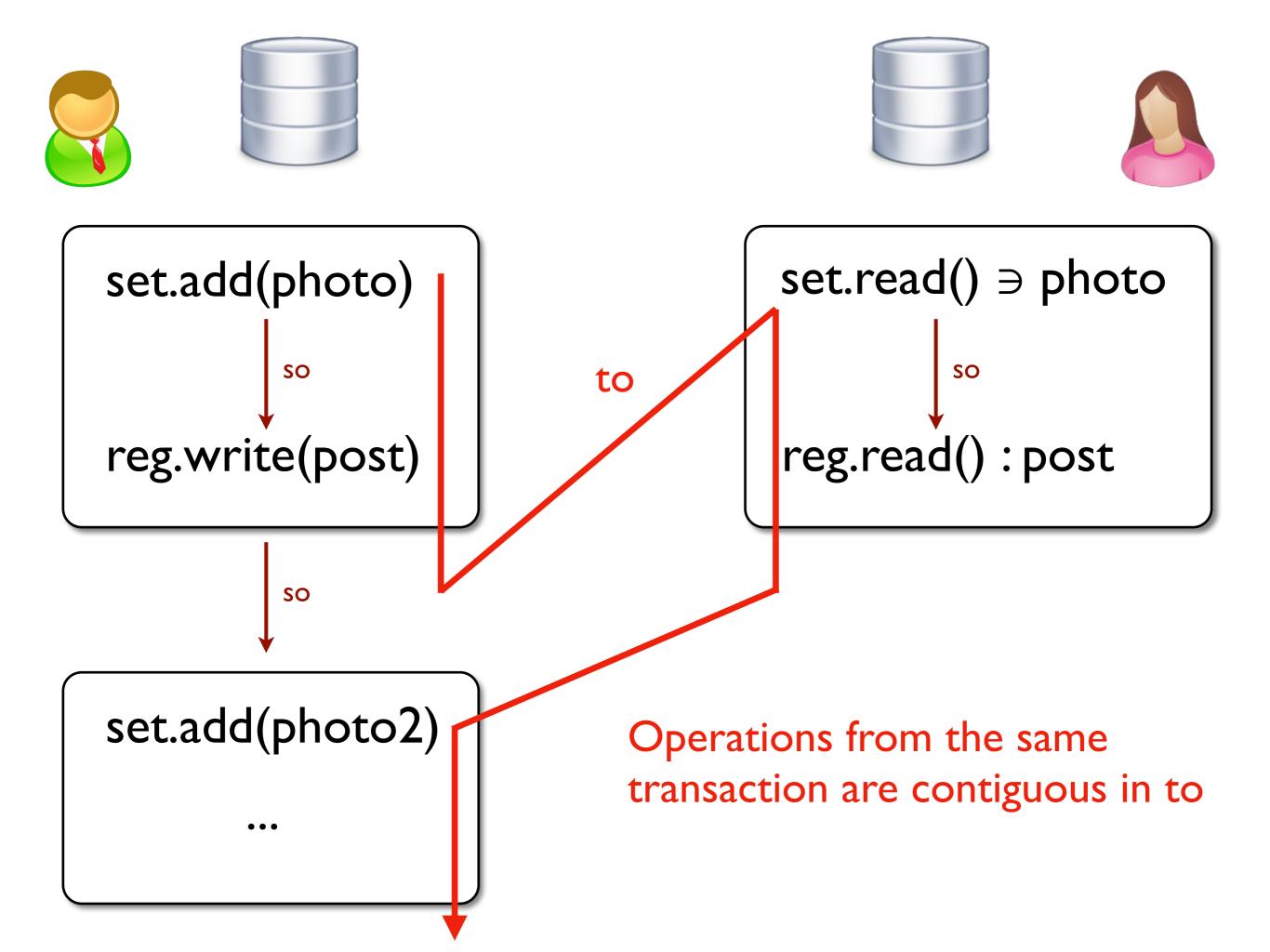
Serializability

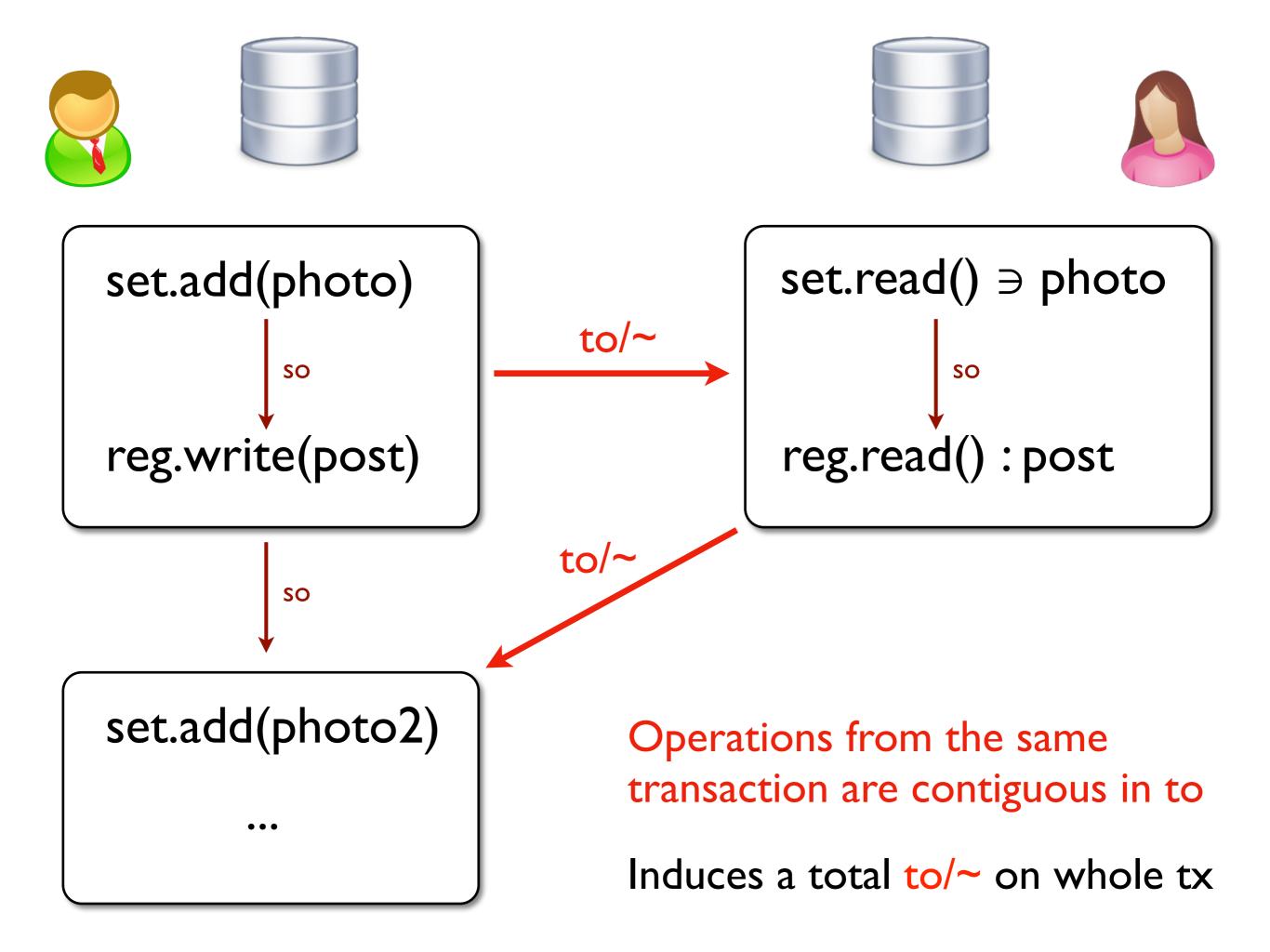
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- 2. The return value of each operation in E is computed from a state obtained by executing all operations on the same object preceding it in to
- 3. Operations from the same transaction are contiguous in to





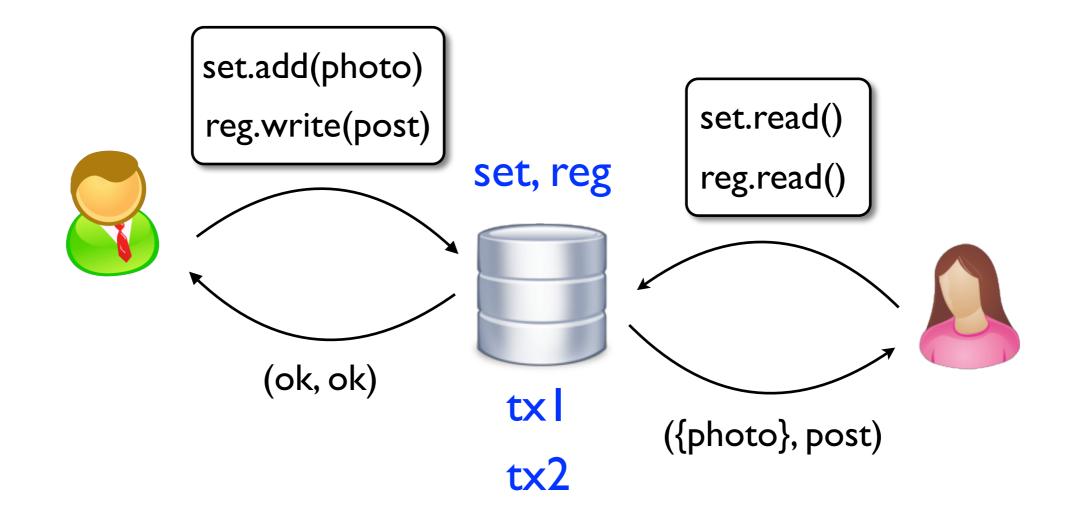


Weakening consistency

• Even single-node databases don't provide serializability either by default or at all: read committed, snapshot isolation, ...

Weakening consistency

- Even single-node databases don't provide serializability either by default or at all: read committed, snapshot isolation, ...
- To better exploit single-node parallelism



Eventually consistent transactions

- Single-node consistency models also applicable in distributed setting
- But many still require some synchronisation between replicas: unavailability, high latency
- Want eventually consistent transactions: always available, low latency
- Preserve some aspects of the invisibility abstraction

System model recap

- Database system consisting of multiple reliable replicas
- Each replica stores a copy of all objects of replicated data types
- Replicas can communicate via asynchronous reliable channels



x.write(post) y.write(comment) x.read : post

- A client connects to a replica and issues transactions
- High availability: the transaction commits immediately, without communication with other replicas, no aborts!



x.write(post) y.write(comment) x.read : post

x.read : post y.read : comment

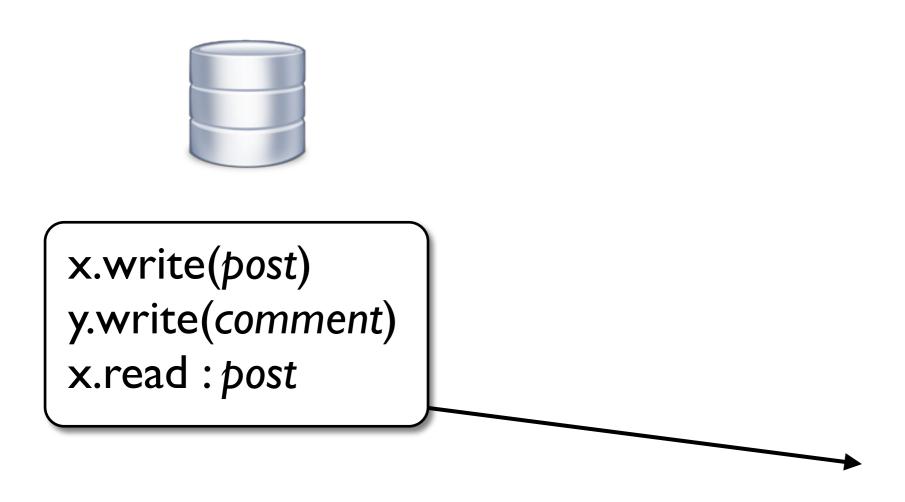
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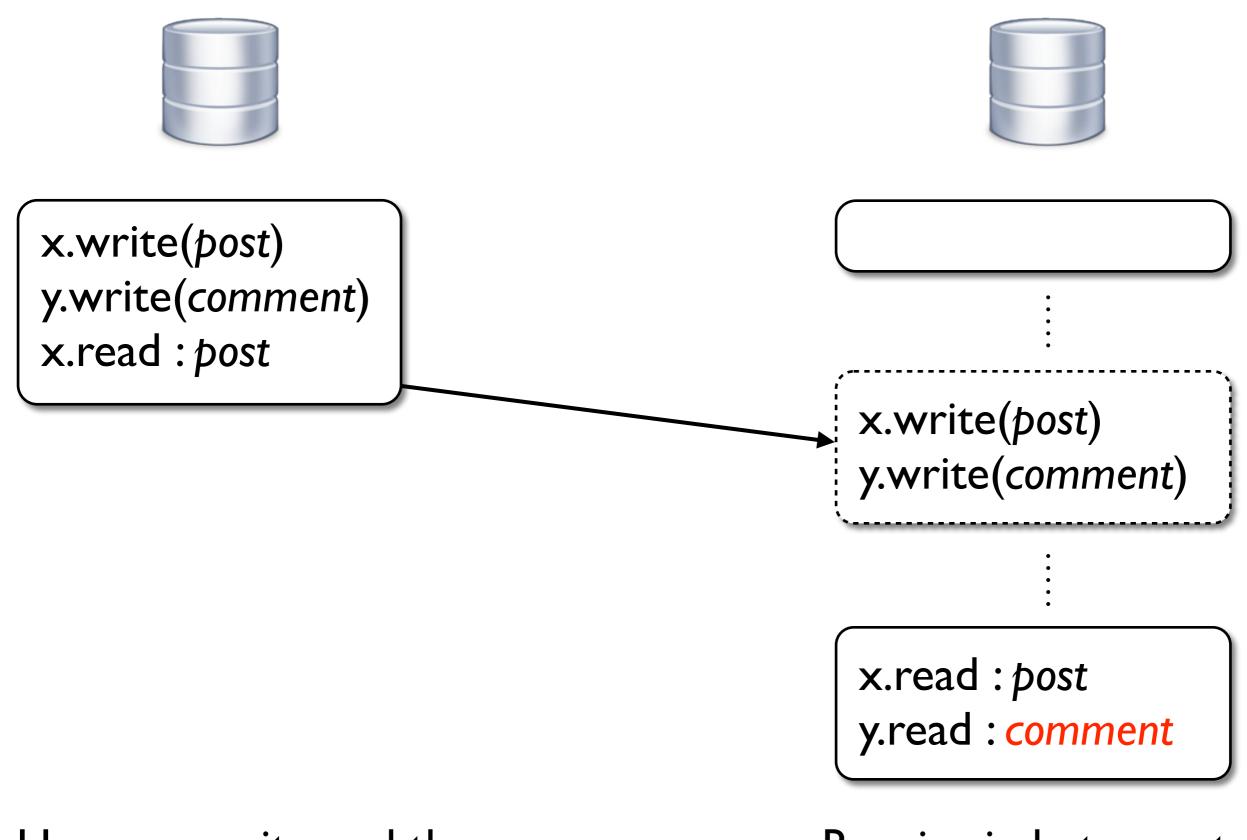
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- A client connects to a replica and issues transactions
- High availability: the transaction commits immediately, without communication with other replicas, no aborts!
- Replica processes transactions sequentially: anomalies arising from single-node concurrency covered by the absence of inter-node synchronisation
- Reads are indivisible: access a fixed snapshot of the database (plus own writes)

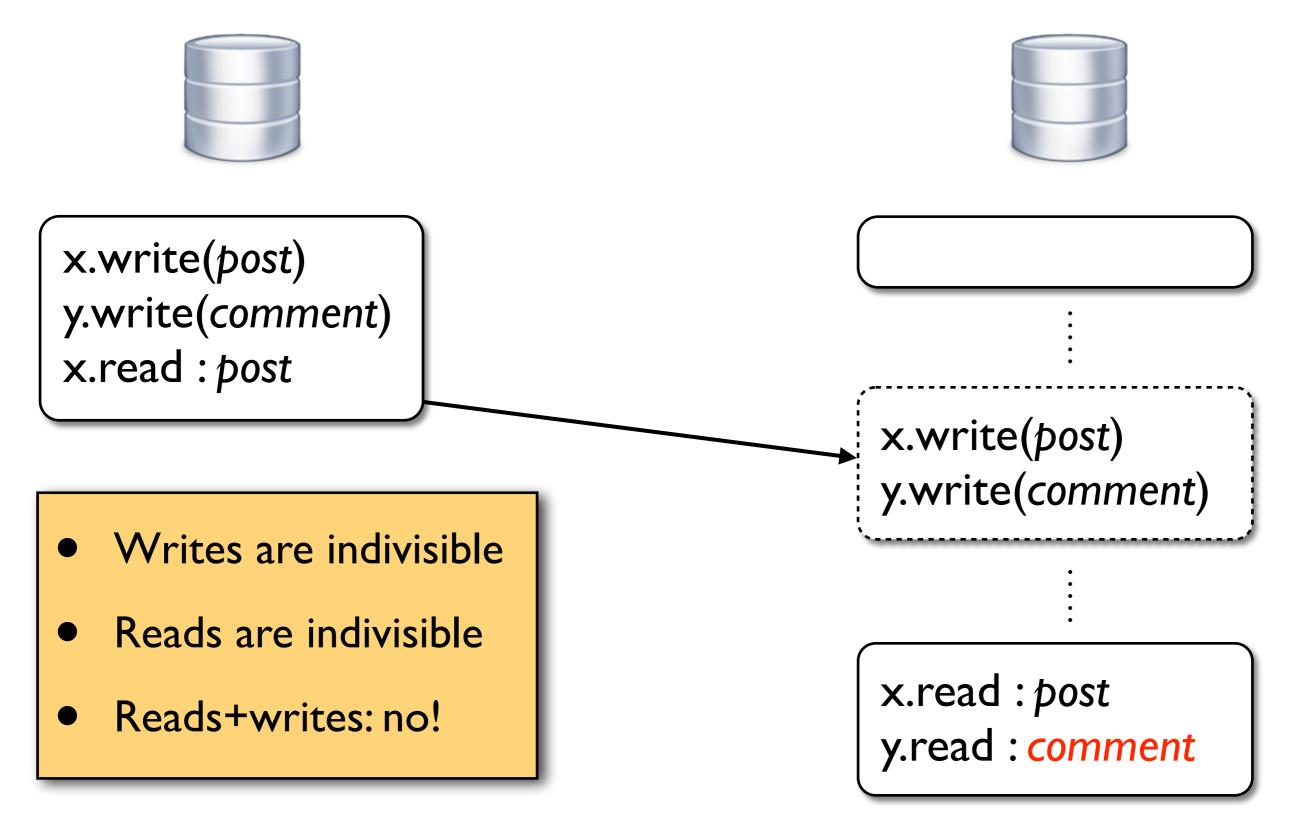


Upon commit: send the effectors of all tx operations to other replicas together



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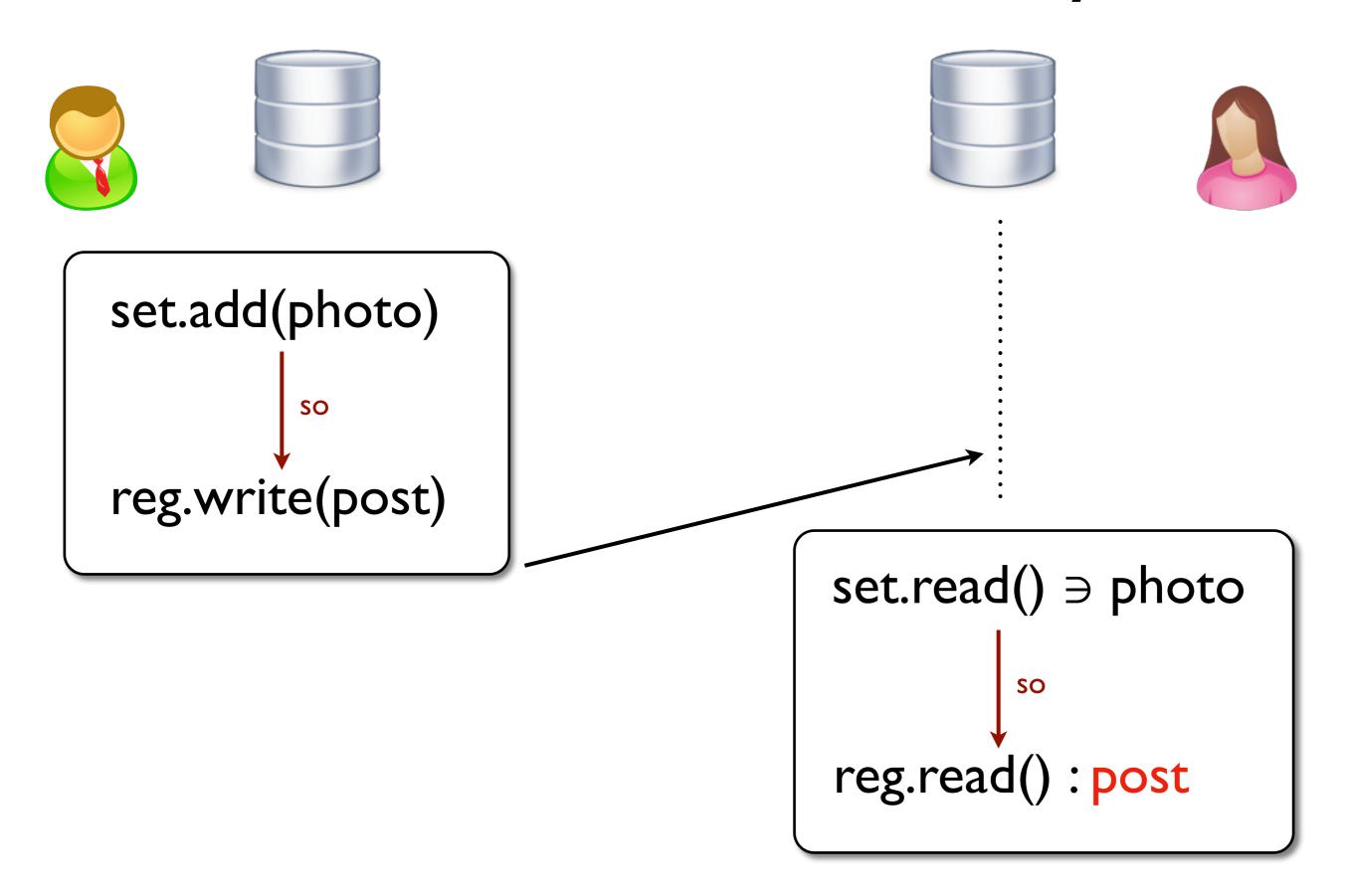
Receive in between txs: incorporate all the updates together



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Receive in between txs: incorporate all the updates together

Reads/writes indivisibility

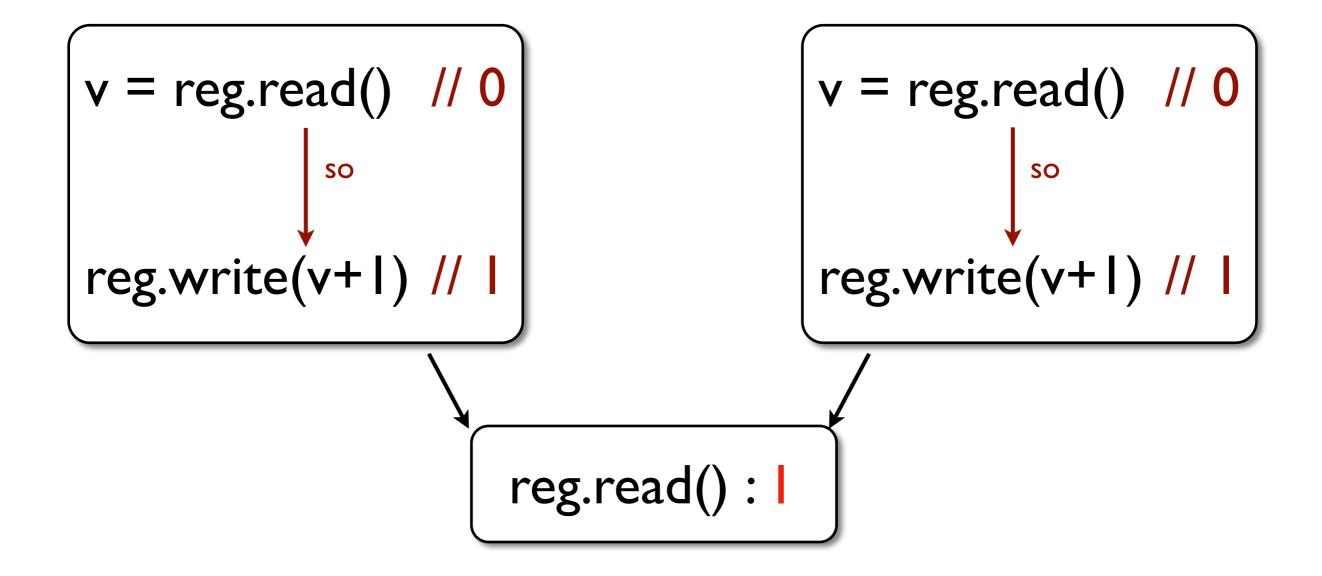


No reads+writes indivisibility

reg: last-writer-wins register, initially 0

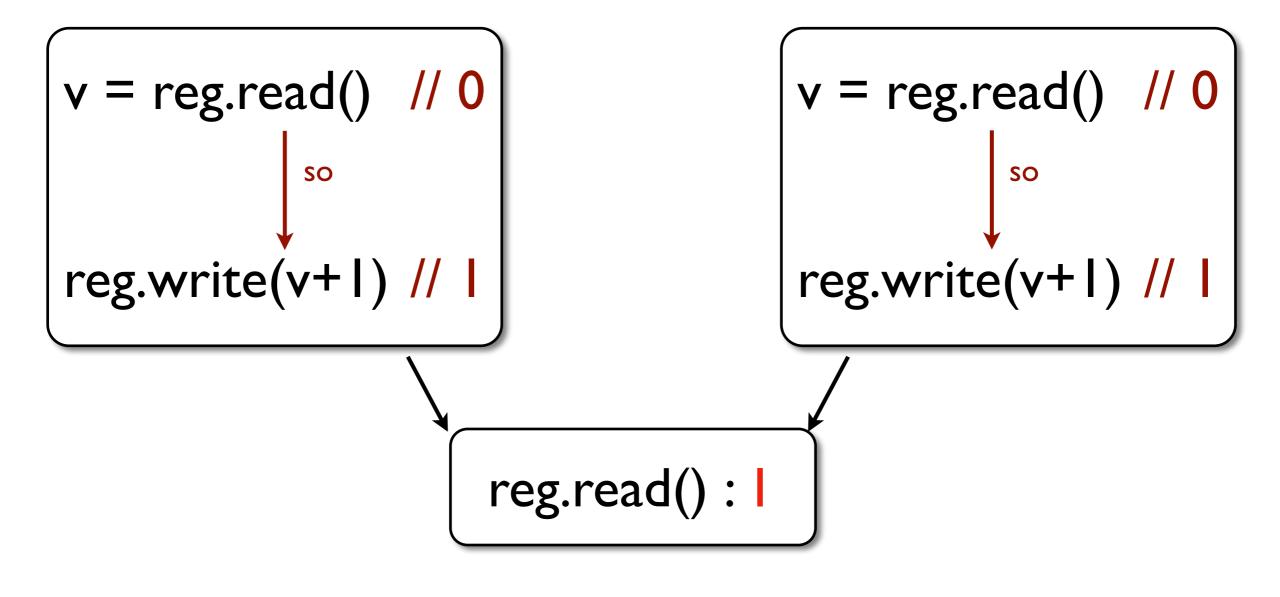
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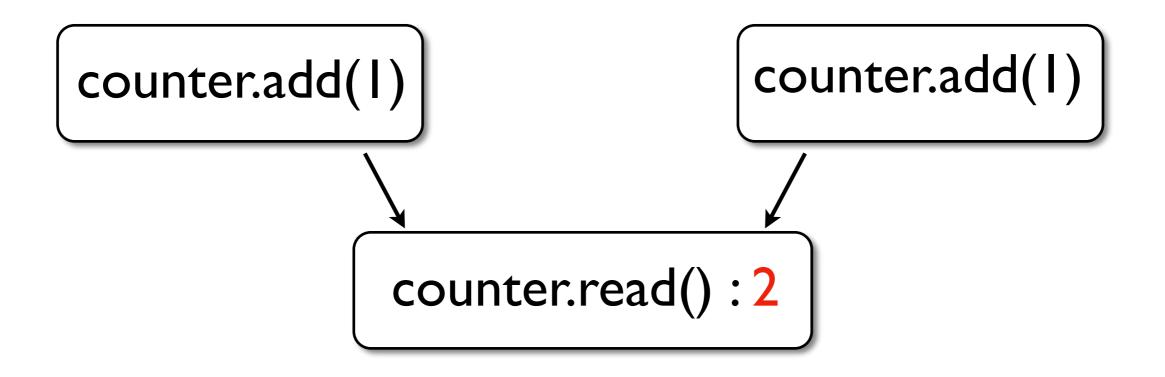
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Lost update anomaly

Use appropriate data type

counter: replicated counter, accumulates increments initially 0

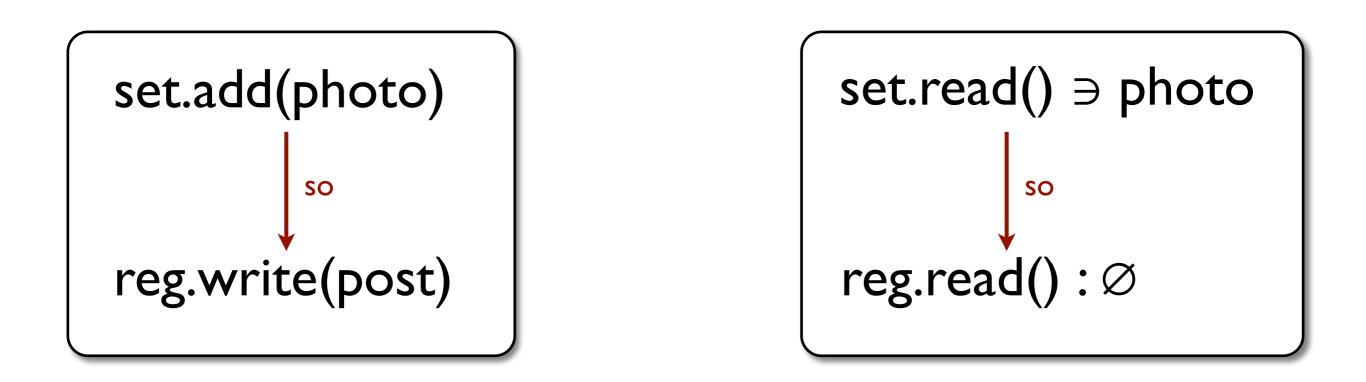


Operational specification

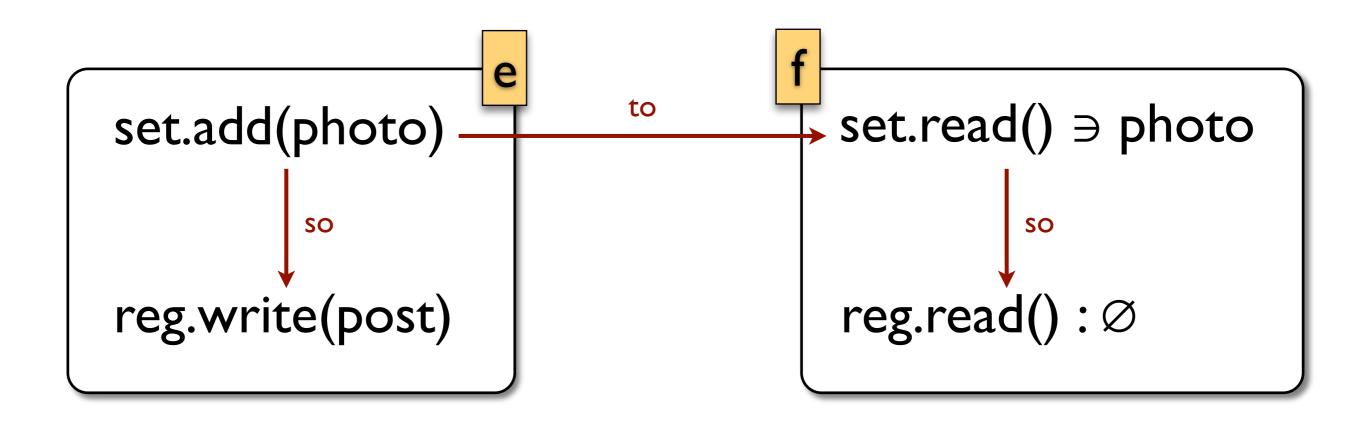
- Eventual consistency with transactions = the set of all histories produced by arbitrary client interactions with the data type implementations (with any allowed message deliveries)
- Implies quiescent consistency: if no new updates are made to the database, then replicas will eventually converge to the same state

Axiomatic specification

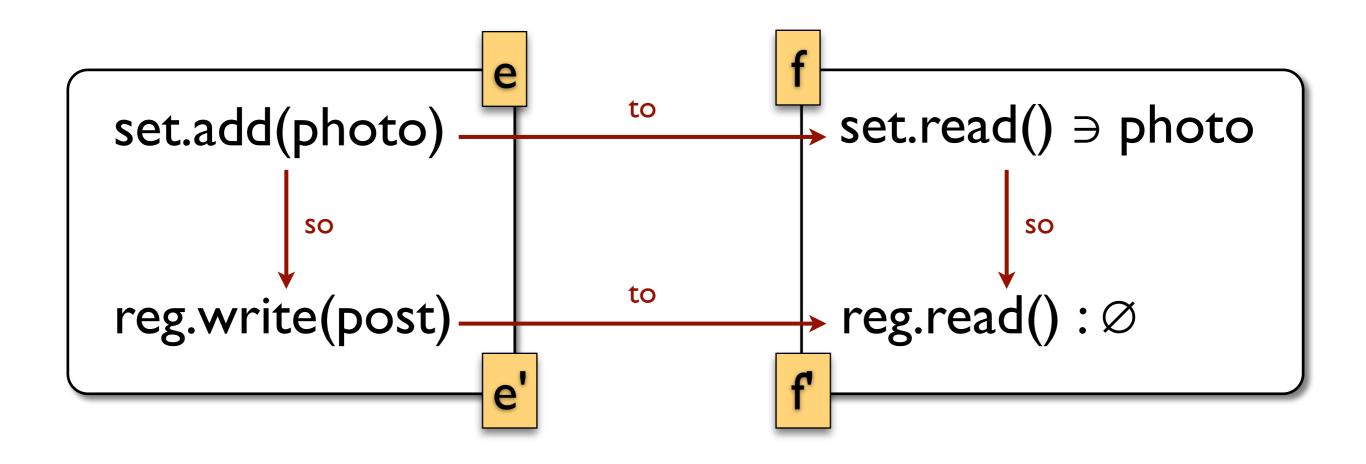
- Serializability: operations from the same transaction are contiguous in the total order to
- Approach: require the same of vis and ar



Operations from the same transaction are contiguous in to: $\forall e, f, e', f'. e \not\sim f \land e' \sim e \xrightarrow{to} f \sim f' \Longrightarrow e' \xrightarrow{to} f'$

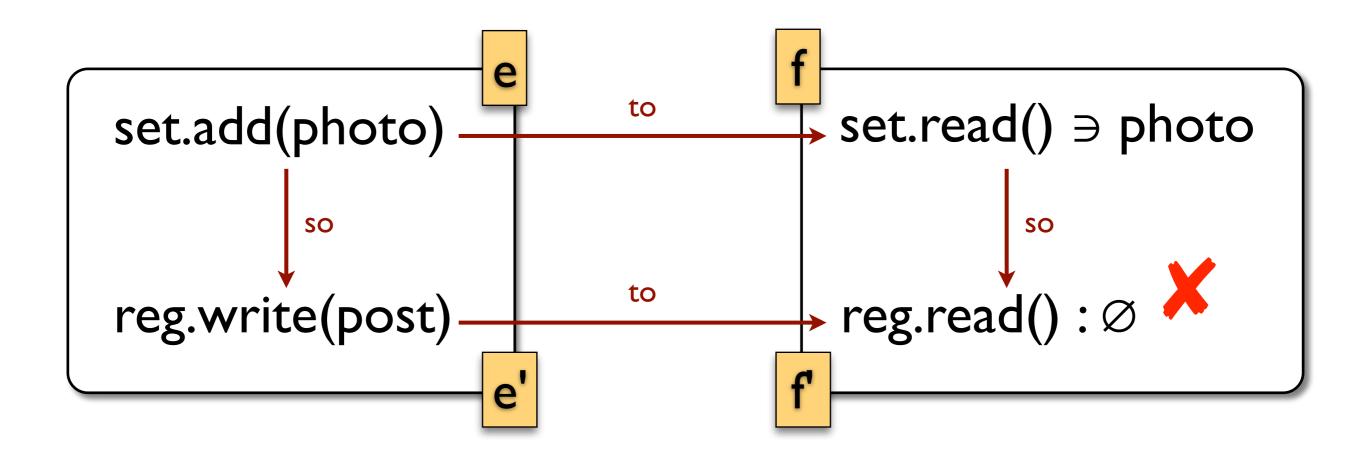


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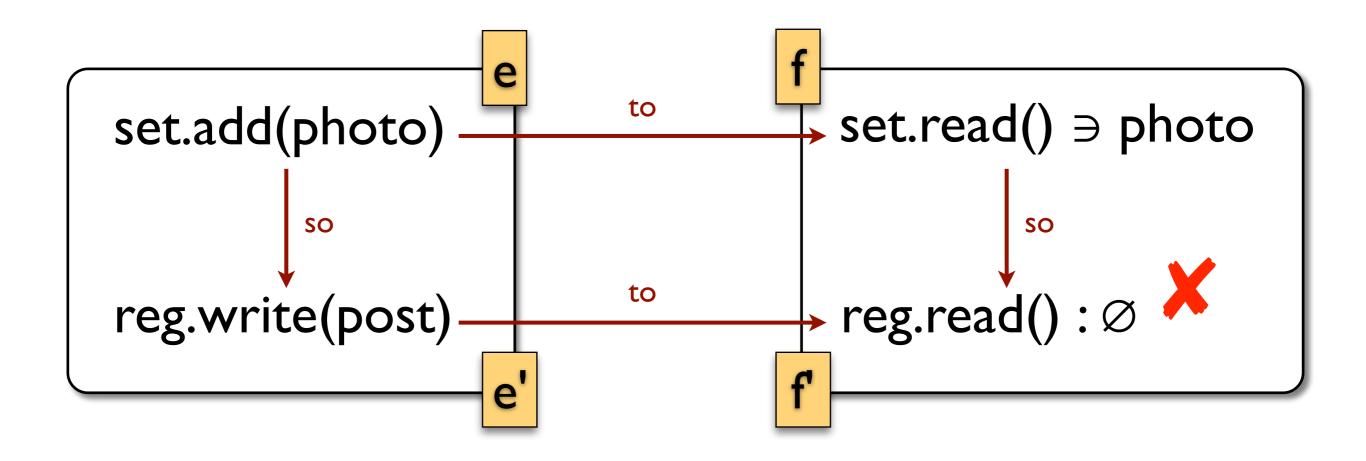
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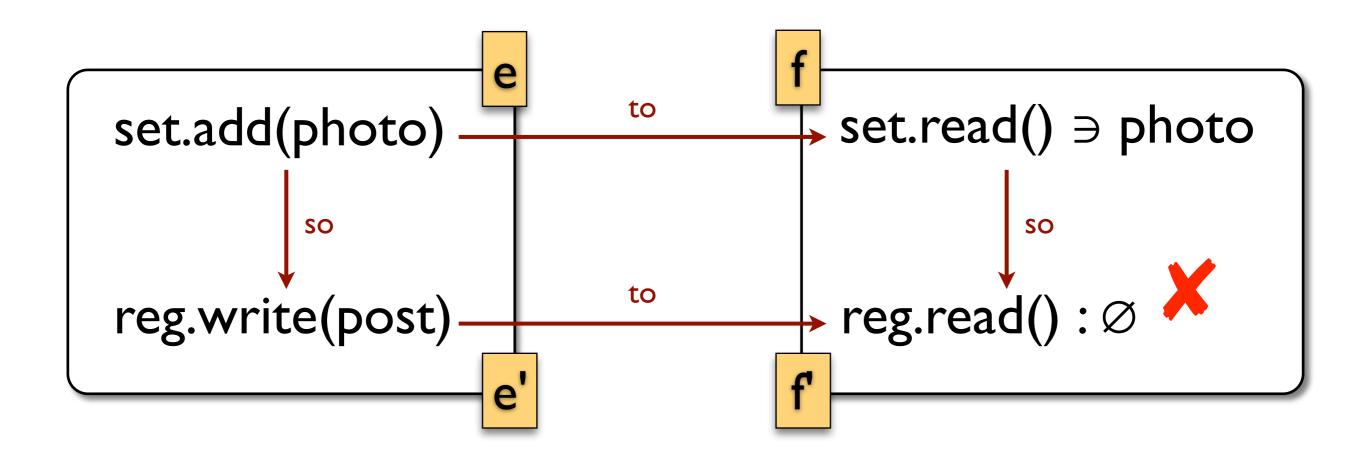
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to treats events in a transaction uniformly

Execution: (E, so, ~, vis, ar)

$$set.add(photo)$$

 so
 $reg.write(post)$
 so
 $reg.read() : \emptyset$

vis, ar treat events in a transaction uniformly: $\forall e, f, e', f'. e \not\sim f \land e' \sim e \xrightarrow{vis} f \sim f' \Longrightarrow e' \xrightarrow{vis} f'$ $\forall e, f, e', f'. e \not\sim f \land e' \sim e \xrightarrow{ar} f \sim f' \Longrightarrow e' \xrightarrow{ar} f'$

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$$\overbrace{set.add(photo) + vis + reg.read() \Rightarrow photo + vis + reg.read() \Rightarrow pho$$

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$$\overbrace{\text{set.add(photo)} + vis} + f$$

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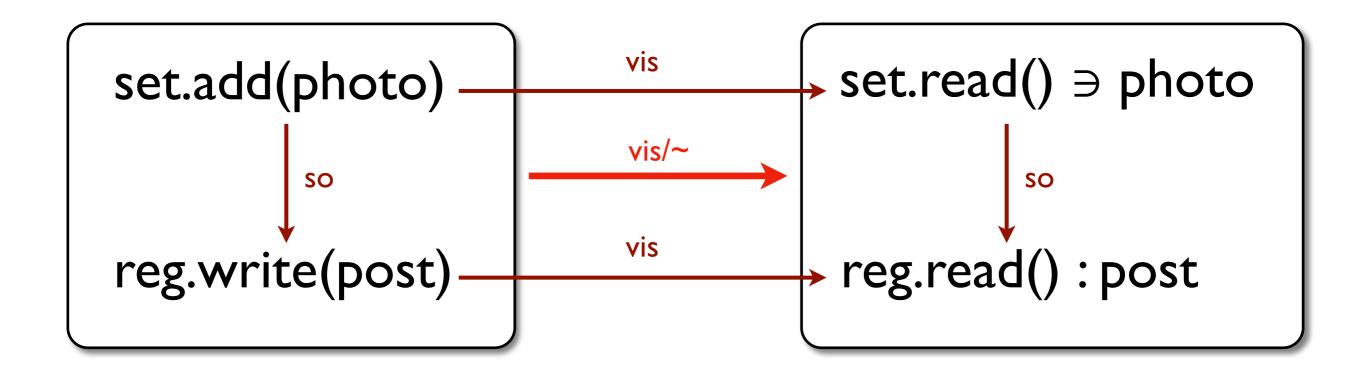
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Execution: (E, so, ~, vis, ar)



vis, ar induce acyclic vis/~, ar/~ on whole txs:

$$T \xrightarrow{\text{vis/}} S \iff \exists e \in T, f \in S. e \xrightarrow{\text{vis}} f$$

$$T \xrightarrow{ar/\sim} S \iff \exists e \in T, f \in S. e \xrightarrow{ar} f$$

Eventually consistent transactions

The set of histories (E, so, \sim) such that for some vis, ar:

- Return values consistent with data type specs: $\forall e \in E. rval(e) = F_{type(obj(e))}(context(e))$
- No causal cycles: so \cup vis is acyclic
- Eventual visibility: $\forall e \in E. e \xrightarrow{vis} f$ for all but finitely many $f \in E$
- Transaction indivisibility: $\forall e, f, e', f'. e \not\sim f \land e' \sim e \xrightarrow{vis} f \sim f' \Longrightarrow e' \xrightarrow{vis} f'$ $\forall e, f, e', f'. e \not\sim f \land e' \sim e \xrightarrow{ar} f \sim f' \Longrightarrow e' \xrightarrow{ar} f'$

Define transactional variants of other consistency models by just adding prior axioms

Serializability: vis = ar

sistent transactions

so, ~) such that for some vis, ar:

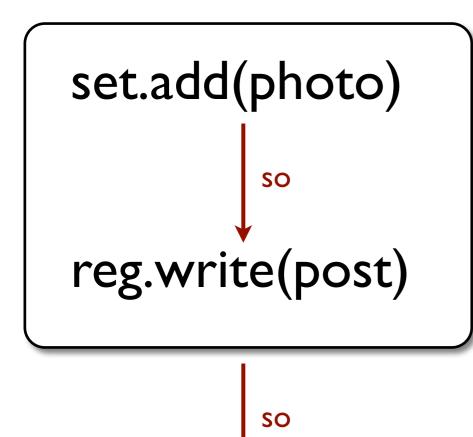
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Session guarantees





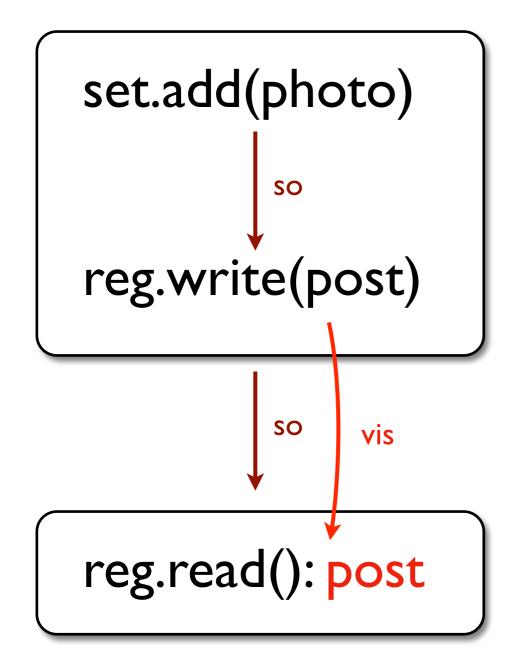
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Transactions in the same session only accumulate information

reg.read(): ?

Session guarantees





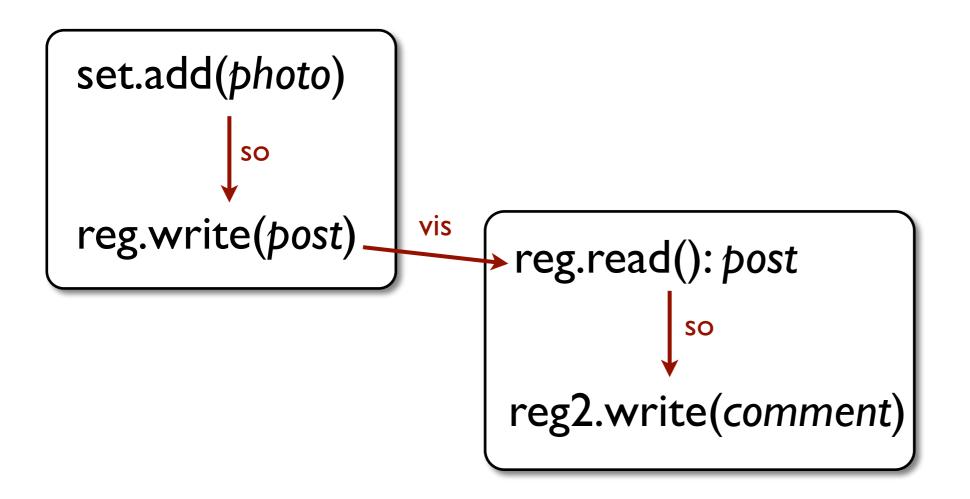
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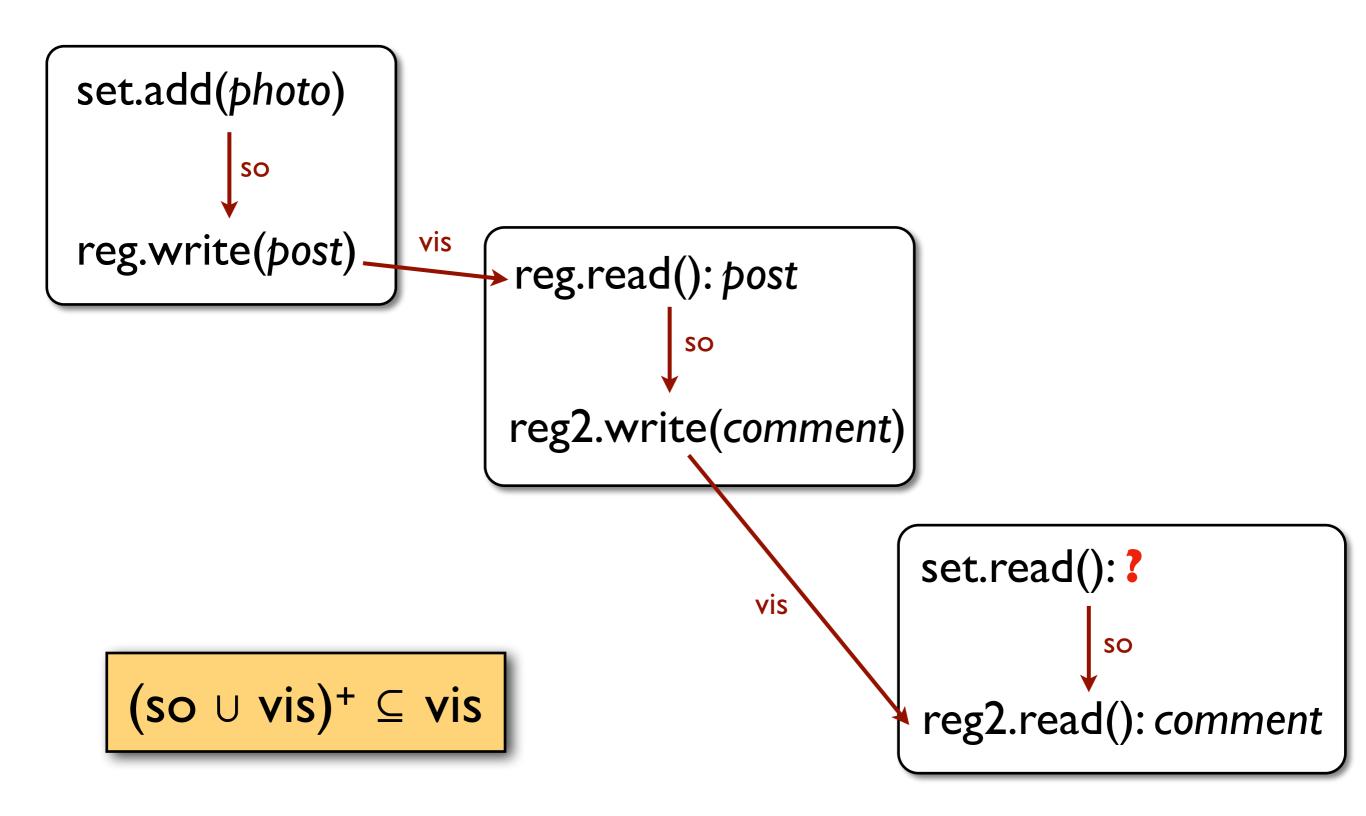
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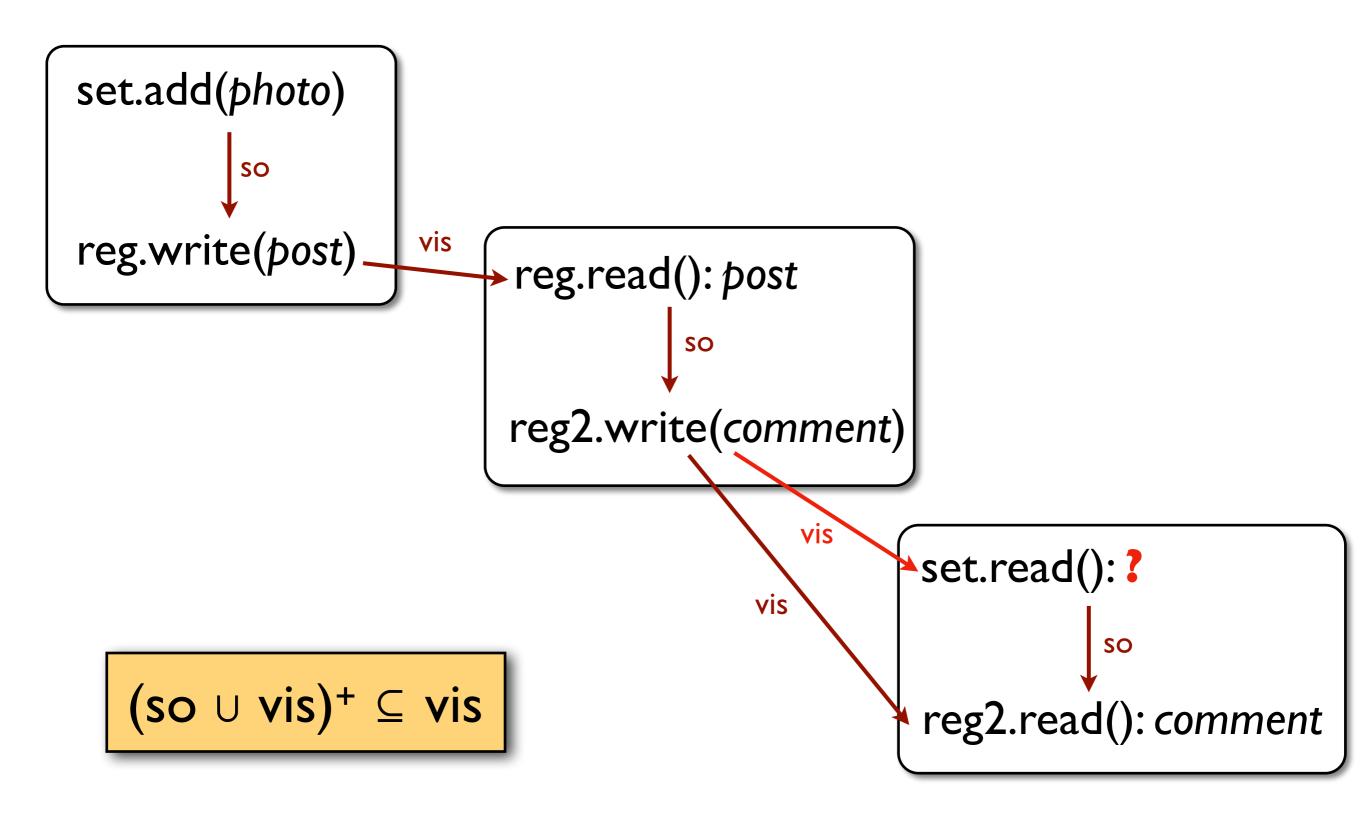


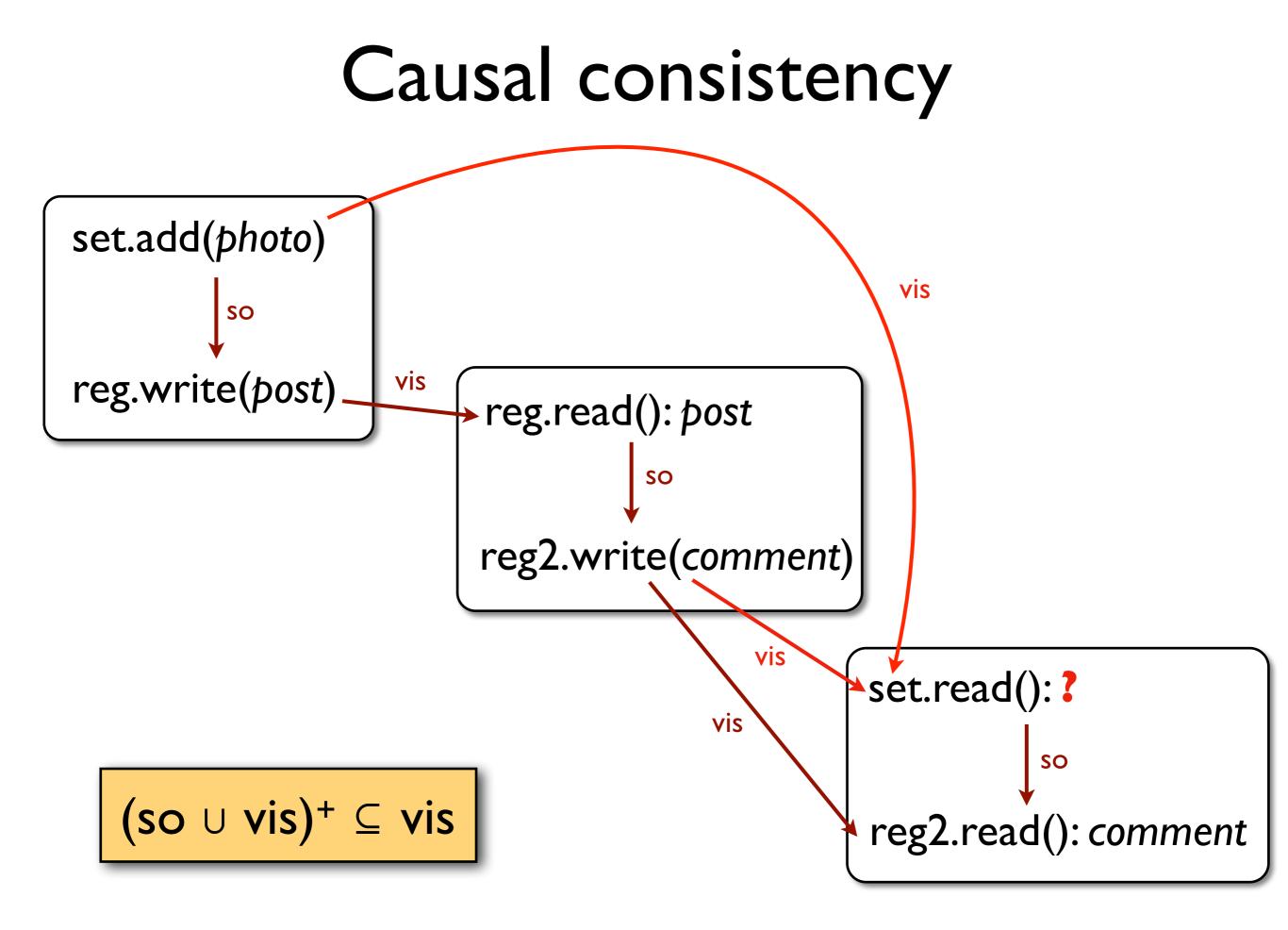
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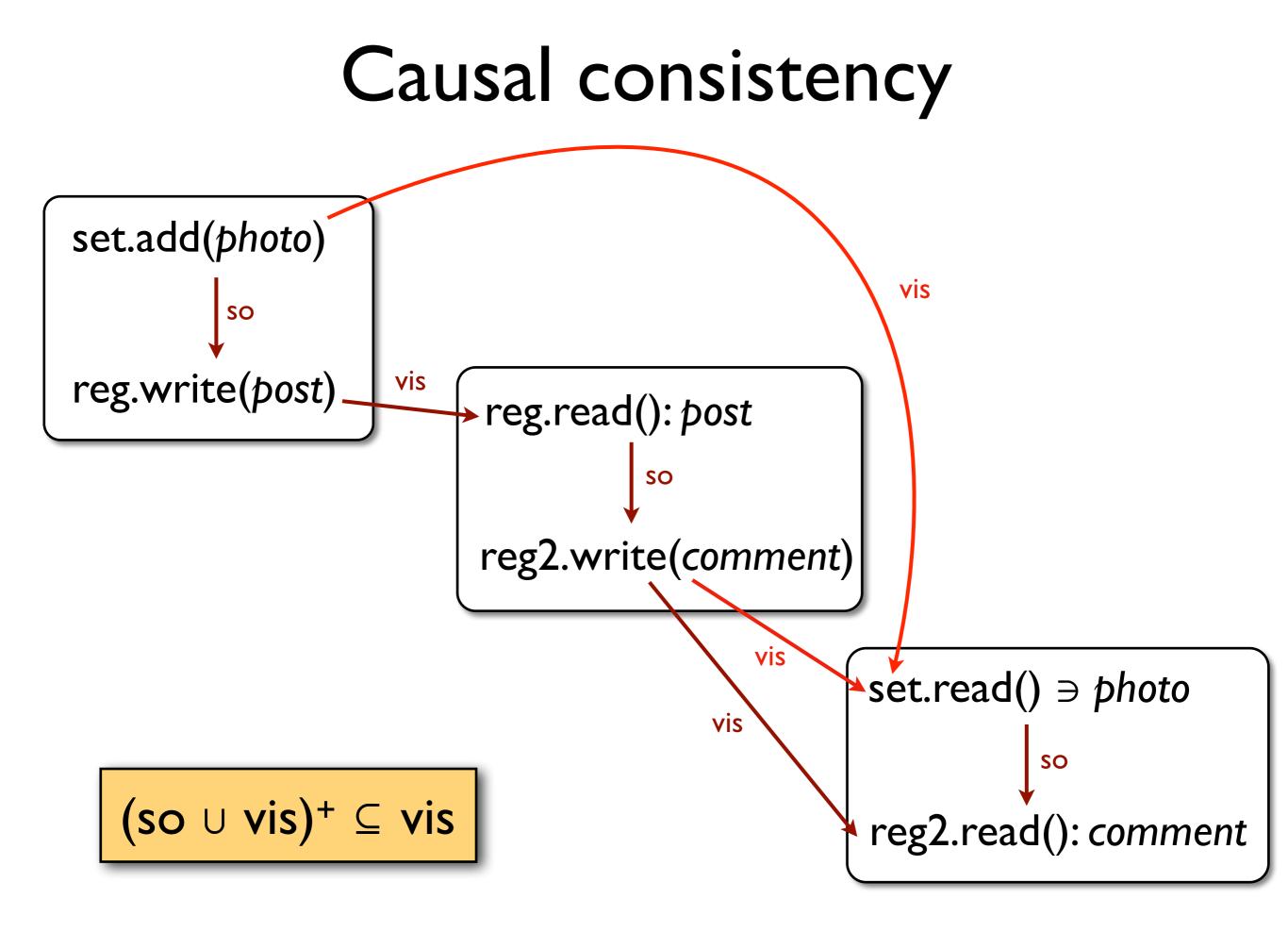


$$(so \cup vis)^+ \subseteq vis$$









Concurrent withdrawals

c: counter with decrements, initially 100

v = c.read()if $(v \ge 100)$ so c.subtract(100)

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if $(v \ge 100)$ so
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Concurrent withdrawals

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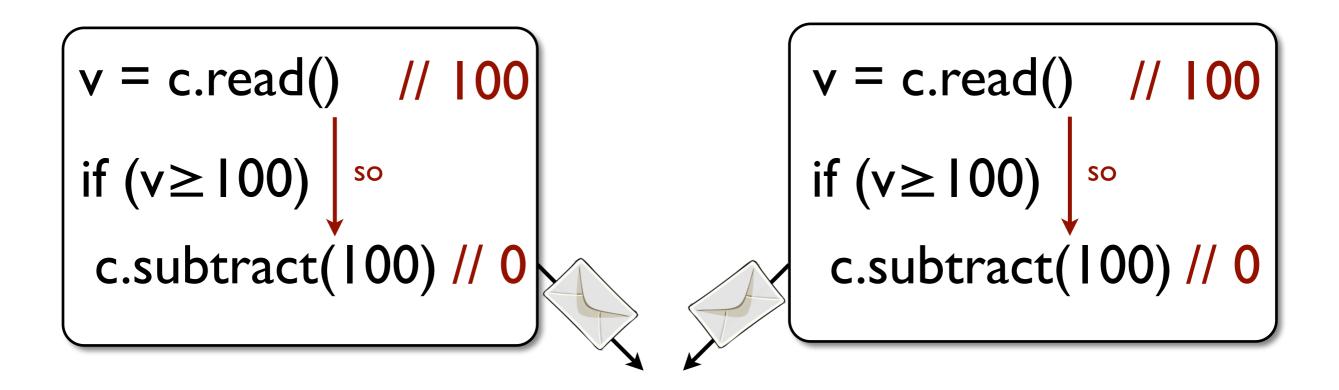
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Concurrent withdrawals

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Both transactions decremented successfully - synchronisation needed!

Recap: strengthening consistency

withdraw(100) :

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- Baseline model: causal consistency
- Symmetric conflict relation on operations: $\bowtie \subseteq \mathsf{Op} \times \mathsf{Op}$, e.g., withdraw \bowtie withdraw
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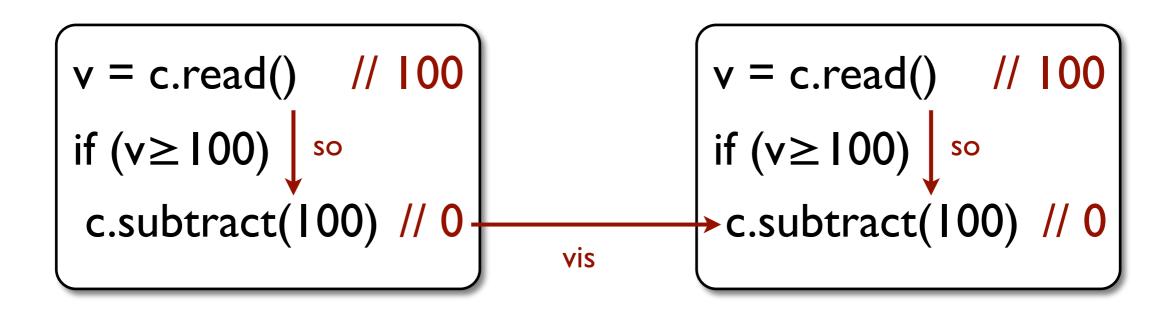
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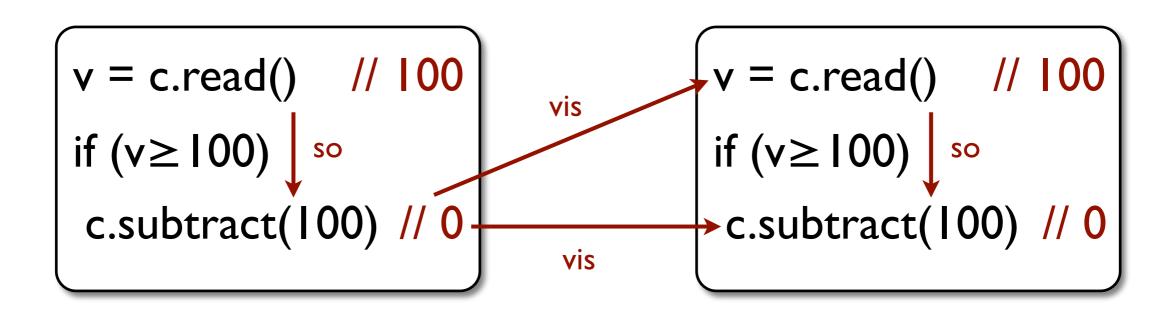
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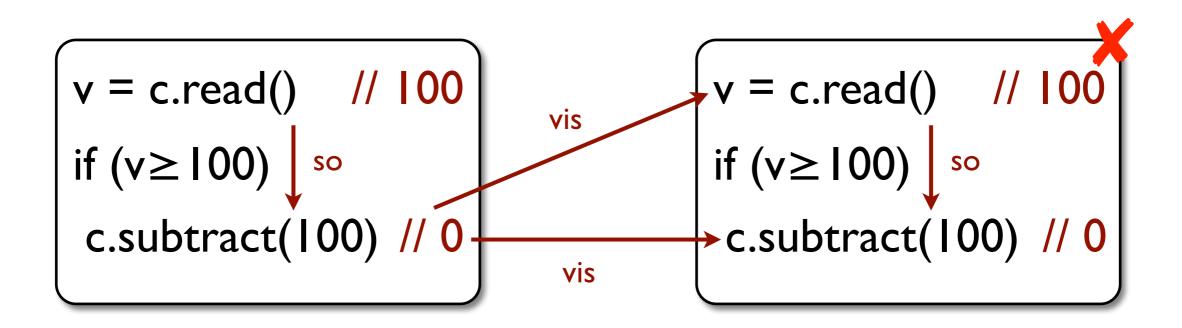
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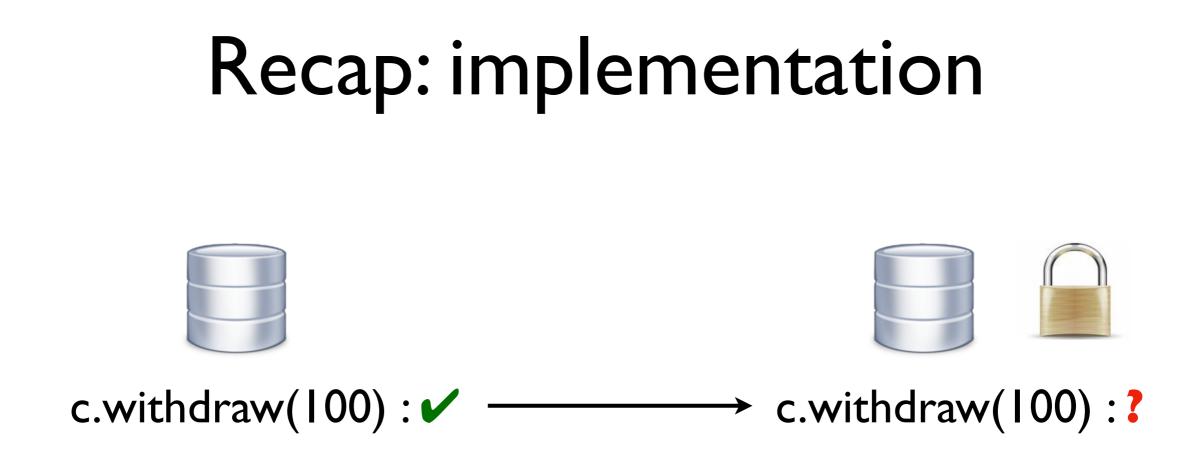
Recap: implementation



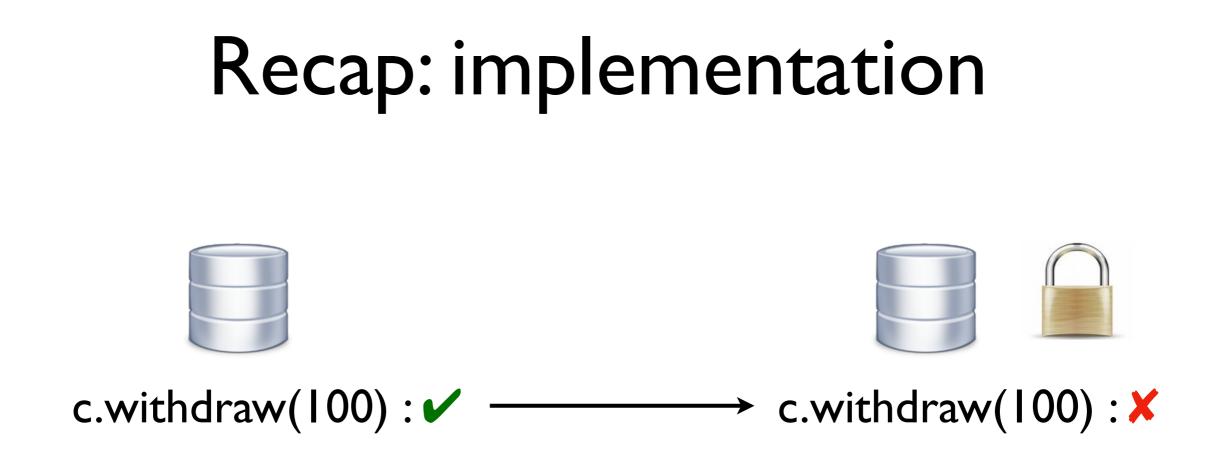


c.withdraw(100) : ?

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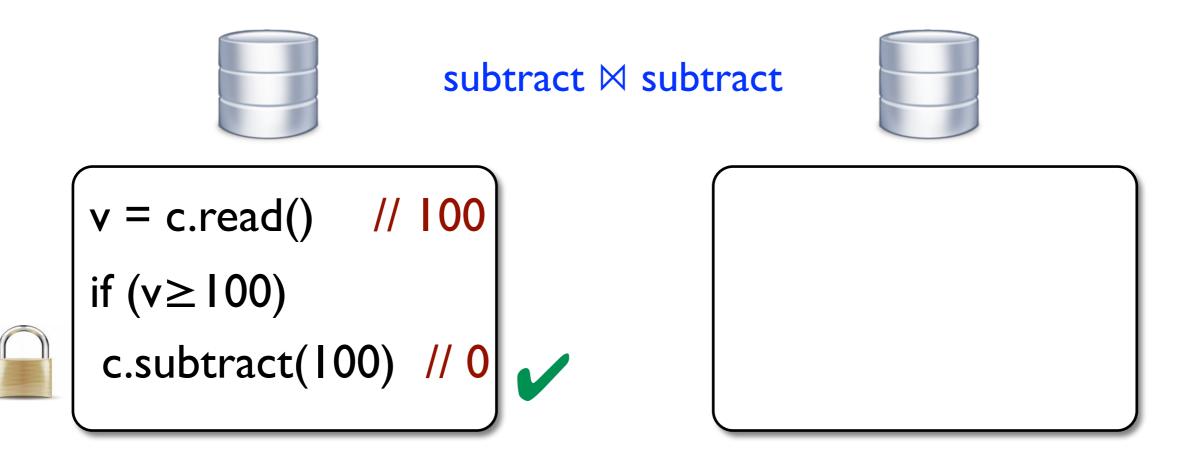
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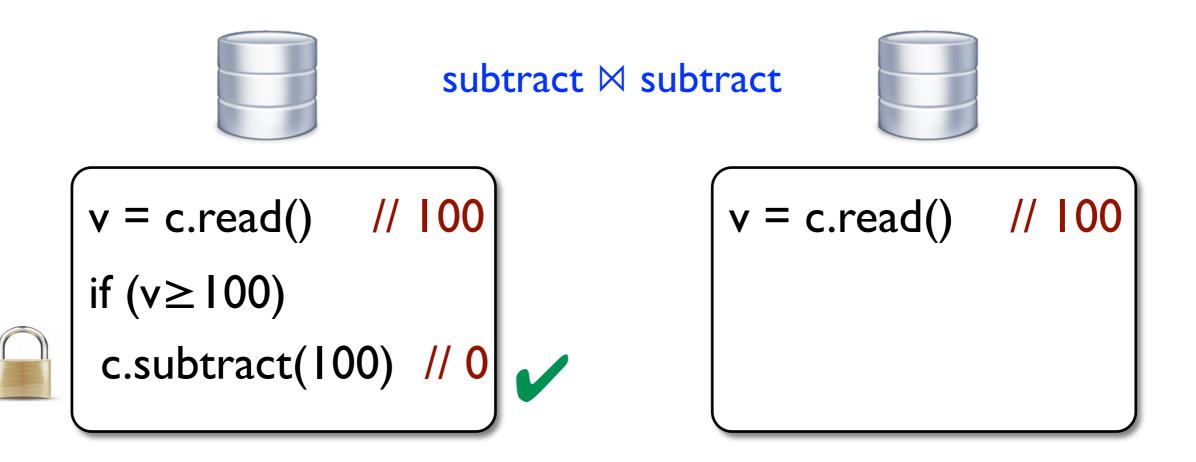


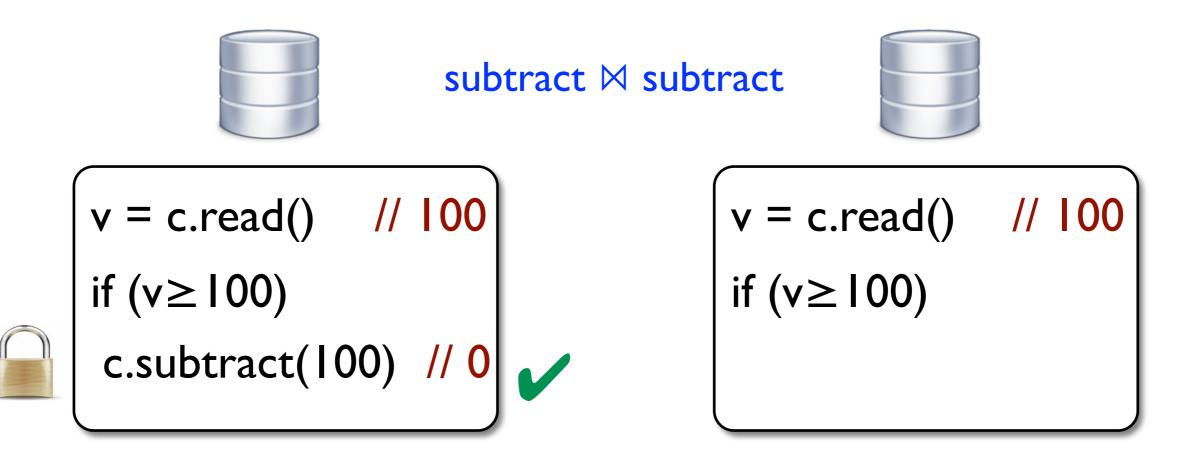
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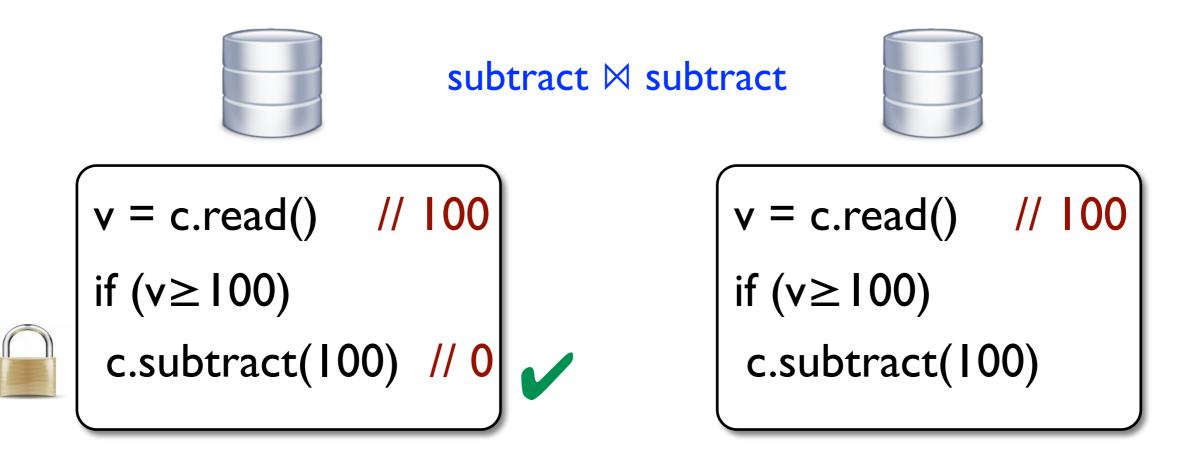
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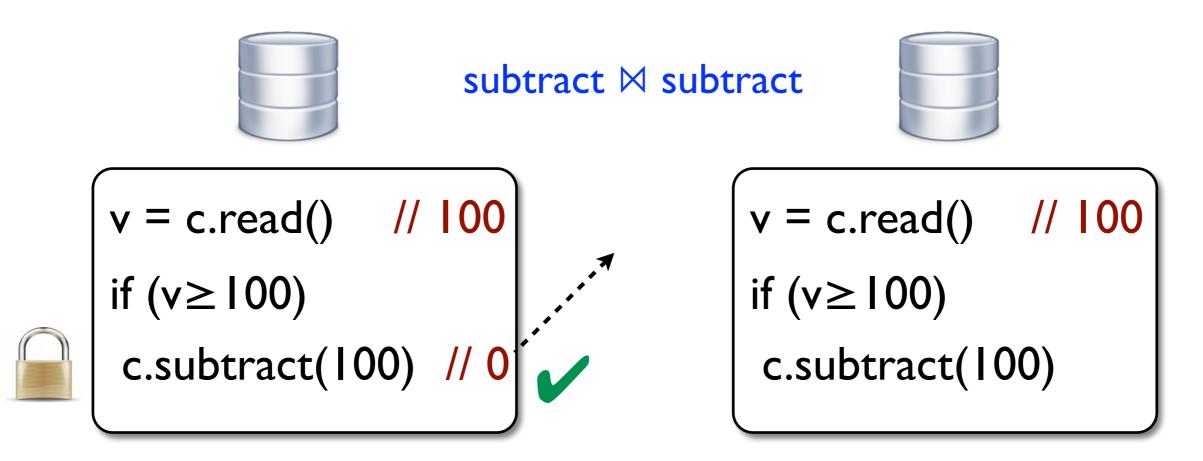
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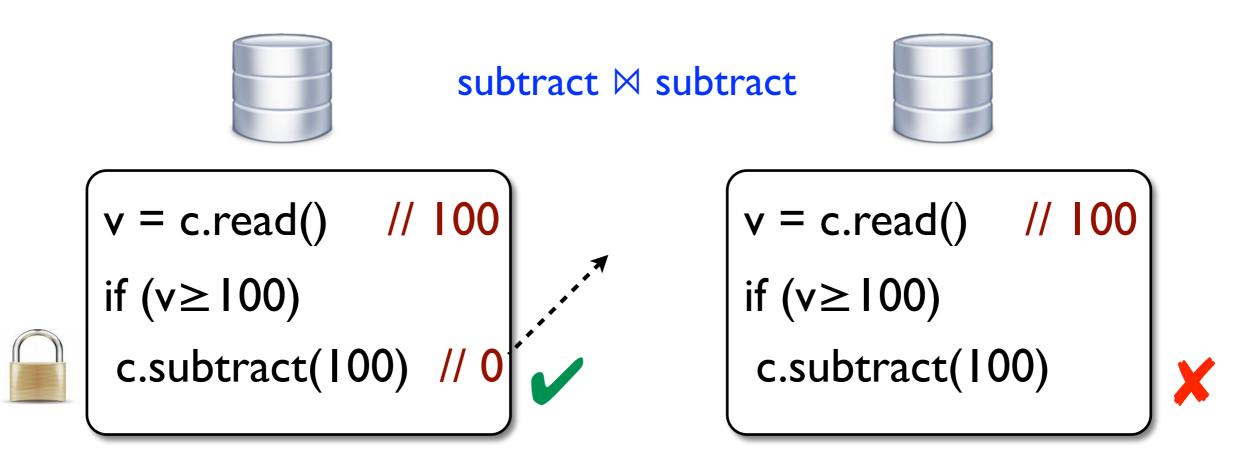




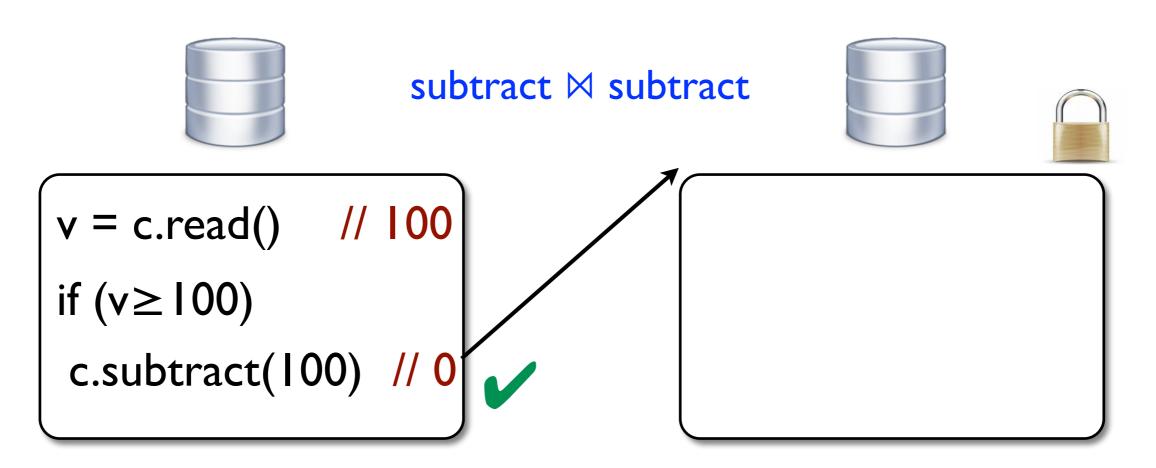




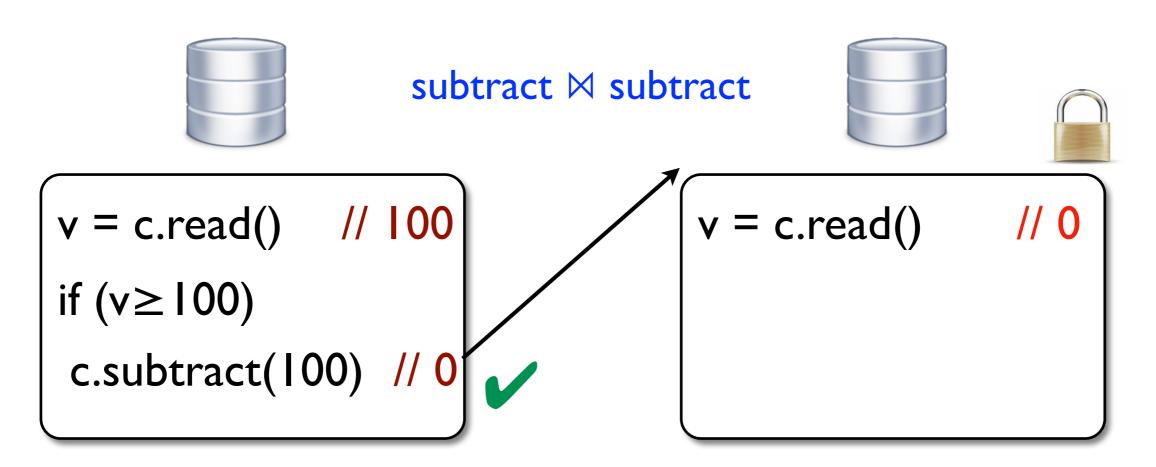
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- Too late: effectors from other replicas only get applied inbetween transactions
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Chosing ⋈

- Want to choose \bowtie to preserve application invariants
- Previous proof rule for checking invariants applies
- Instead of an effector of a single operation, consider a sequential composition of effectors of all operations in a transaction
- Can also fix M so that it's easier to program: new consistency models, disallowing some classes of anomalies

 Operations updating the same object conflict, so cannot be causally independent:

 $\begin{array}{l} \forall e, f \in E. \ obj(e) = obj(f) \land update(op(e)) \land update(op(f)) \\ \Longrightarrow e \stackrel{vis}{\longrightarrow} f \lor f \stackrel{vis}{\longrightarrow} e \end{array} \end{array}$

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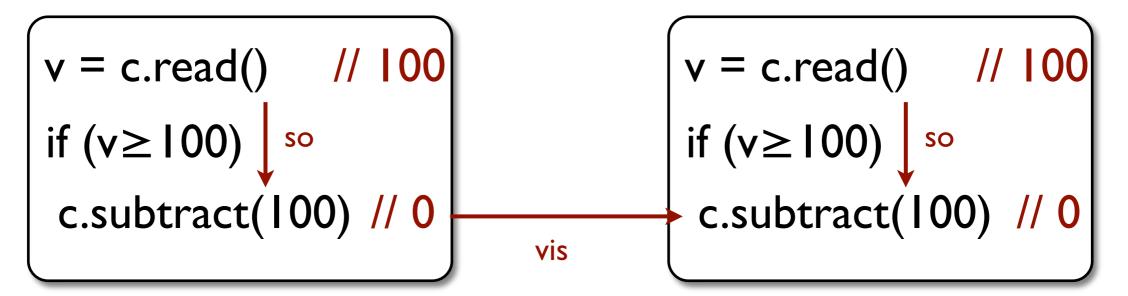
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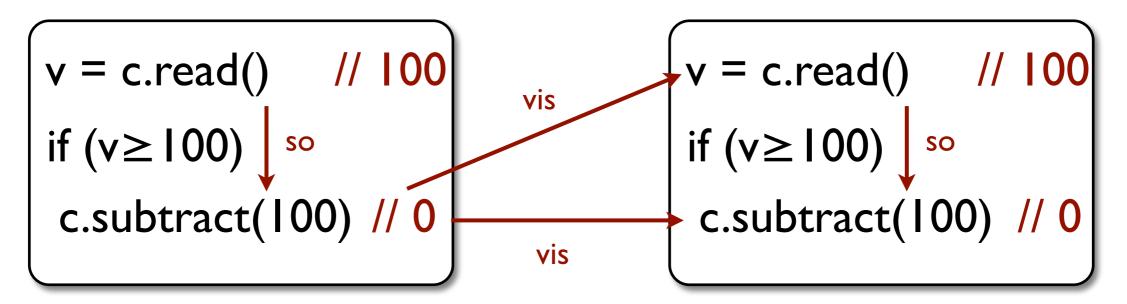
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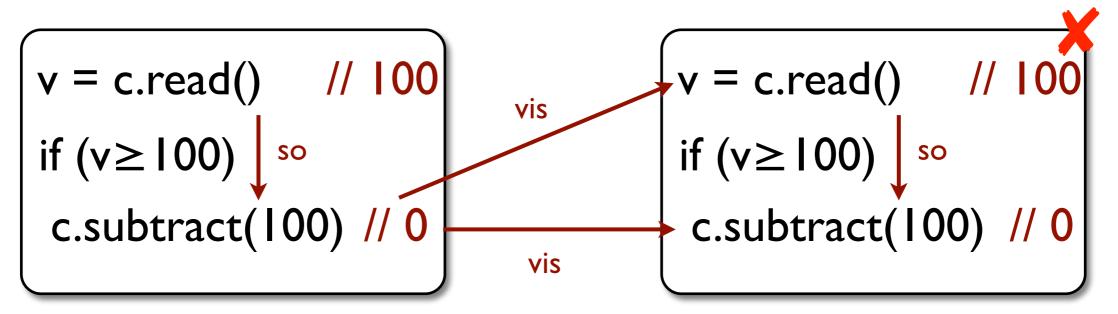
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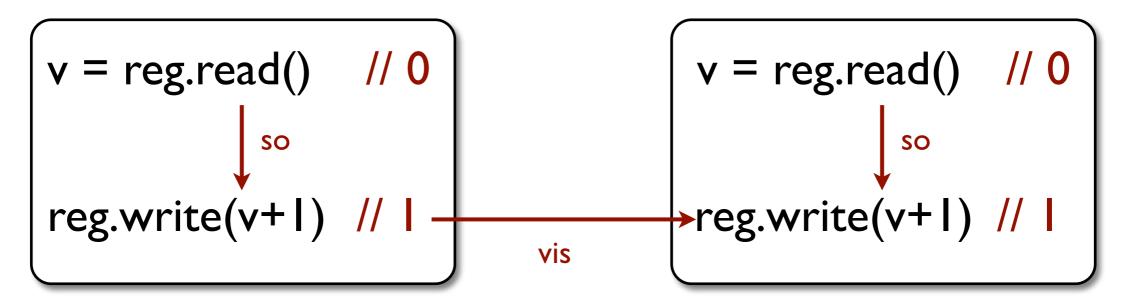


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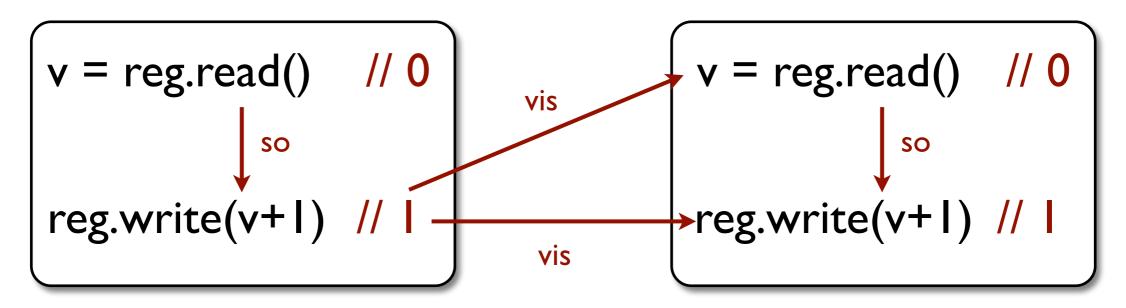
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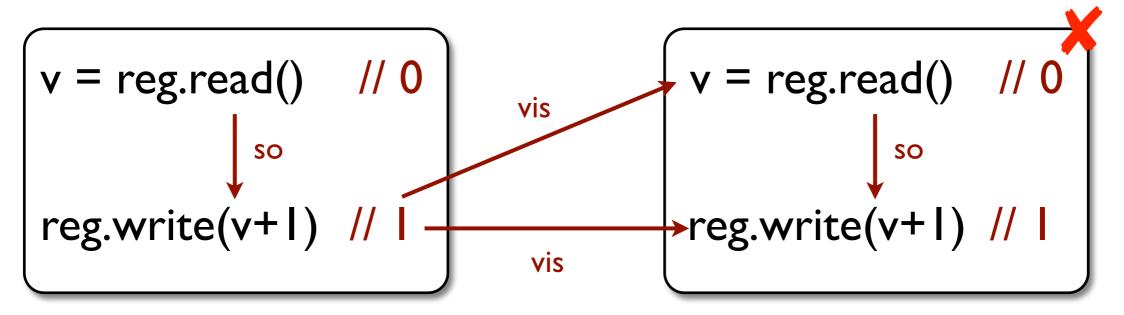
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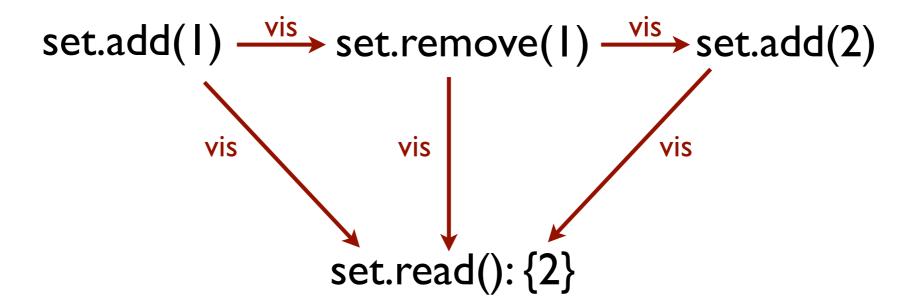
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• Updates on different accounts can go in parallel:

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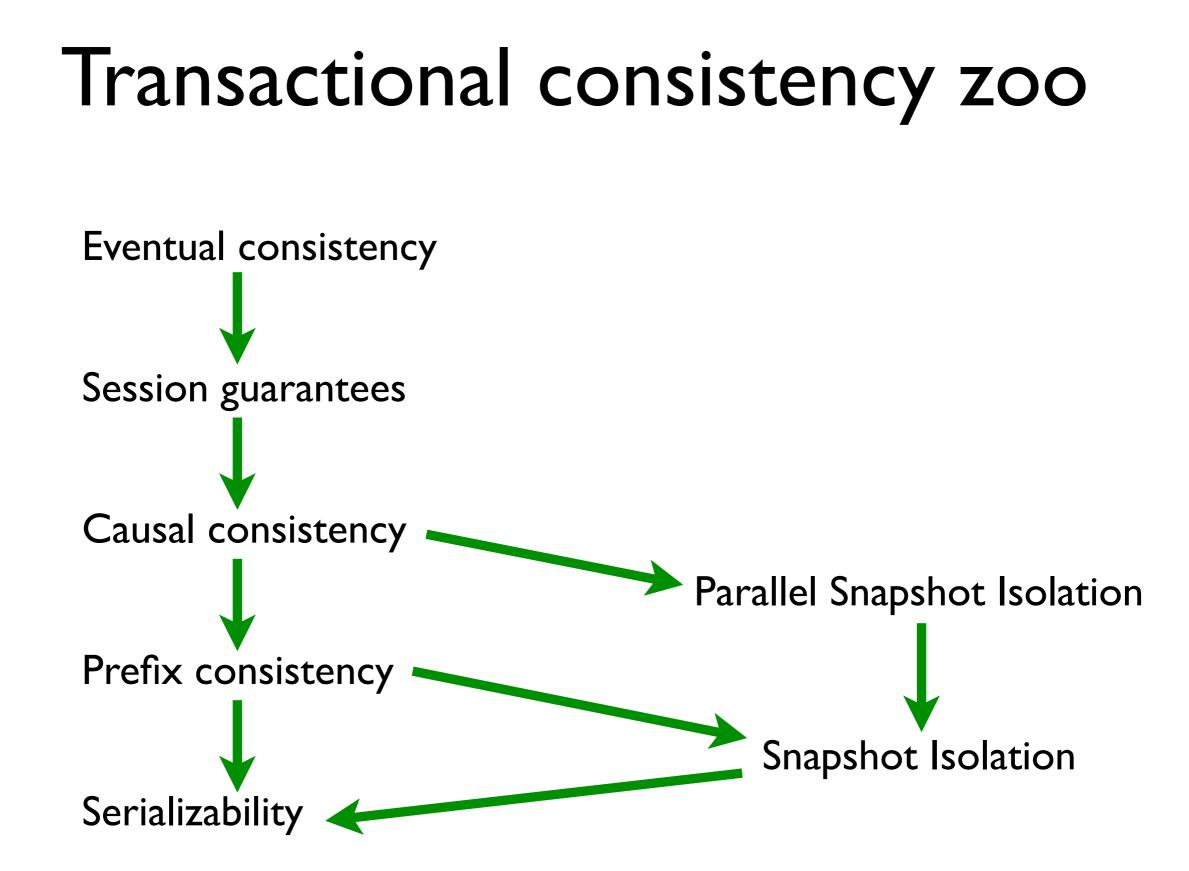
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- Visibility totally orders transactions updating the same object \implies don't need replicated data types, don't need ar
- Can use sequential data types: from now on just sequential read-write registers



Transactional consistency zoo Eventual consistency Session guarantees Causal consistency + write-conflict detection Causal consistency Parallel Snapshot Isolation Prefix consistency **Snapshot Isolation** Serializability

Application correctness

 Does an application satisfy a particular correctness property?

Integrity invariants: account balance is non-negative

Is an application robust against a particular consistency model?

Application behaves the same as when using a strongly consistent database

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Parallel shapshot isolation

• Database with only sequential read-write registers

Assume there is an implicit transaction writing initial values to all registers

PSI = the set of histories (E, so, \sim) such that for some vis:

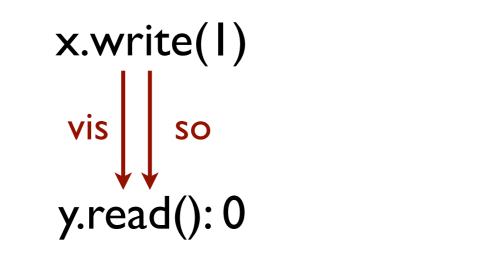
- No causal cycles: so \cup vis is acyclic
- Eventual visibility: $\forall e \in E. e \xrightarrow{vis} f$ for all but finitely many $f \in E$
- Transaction indivisibility: $\forall e, f, e', f'. e \not\sim f \land e' \sim e \xrightarrow{vis} f \sim f' \Longrightarrow e' \xrightarrow{vis} f'$
- Causality preservation: (so \cup vis)⁺ \subseteq vis
- Write-conflict detection: $\forall e, f \in E. obj(e) = obj(f) \land op(e) = write(-) \land op(f) = write(-)$ $\implies e \xrightarrow{vis} f \lor f \xrightarrow{vis} e$
- A read event returns the value written by the last preceding write in vis

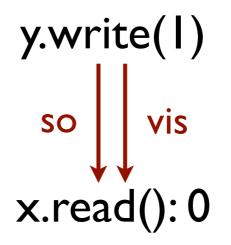
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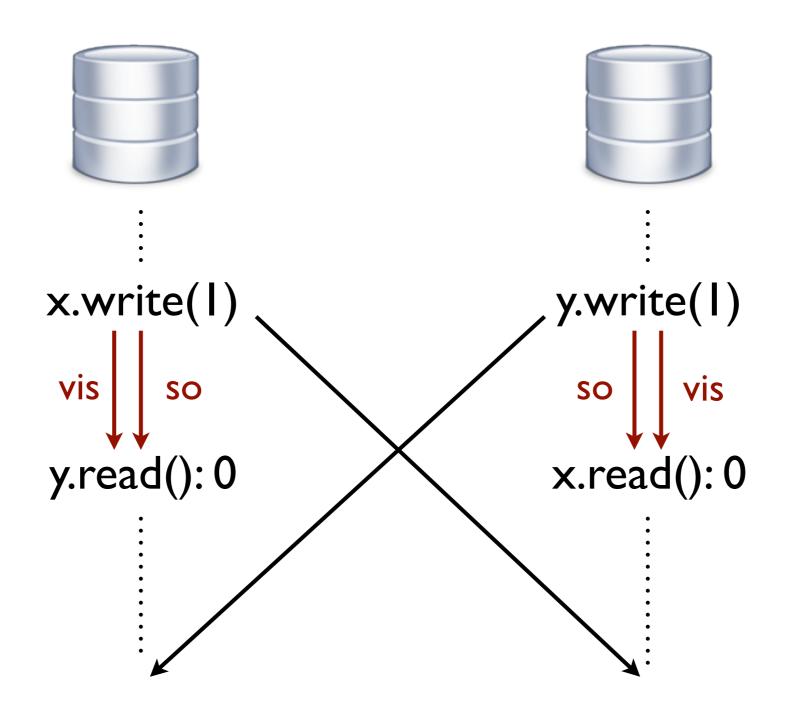
Well-formed because of write-conflict detection

Dekker example

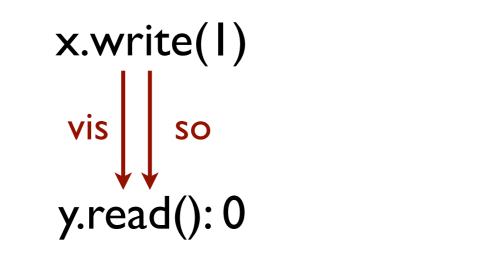


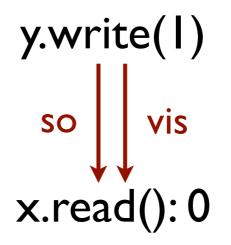


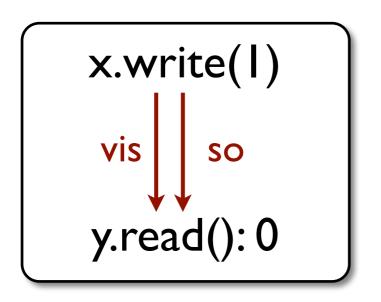
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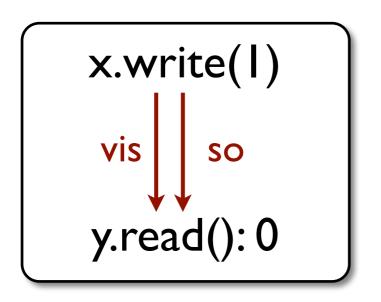


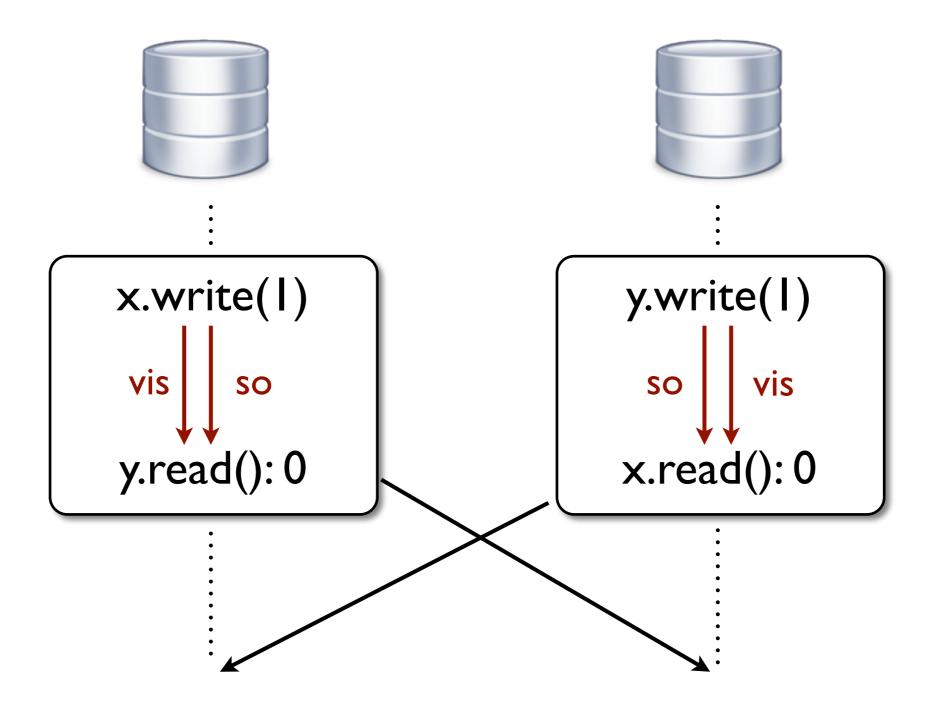




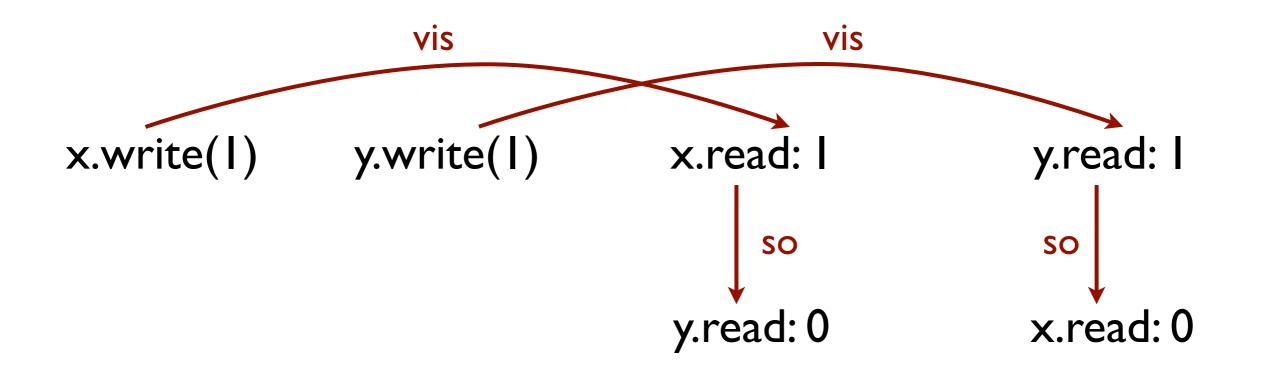


Not serializable, allowed by transactional causal consistency and parallel snapshot isolation

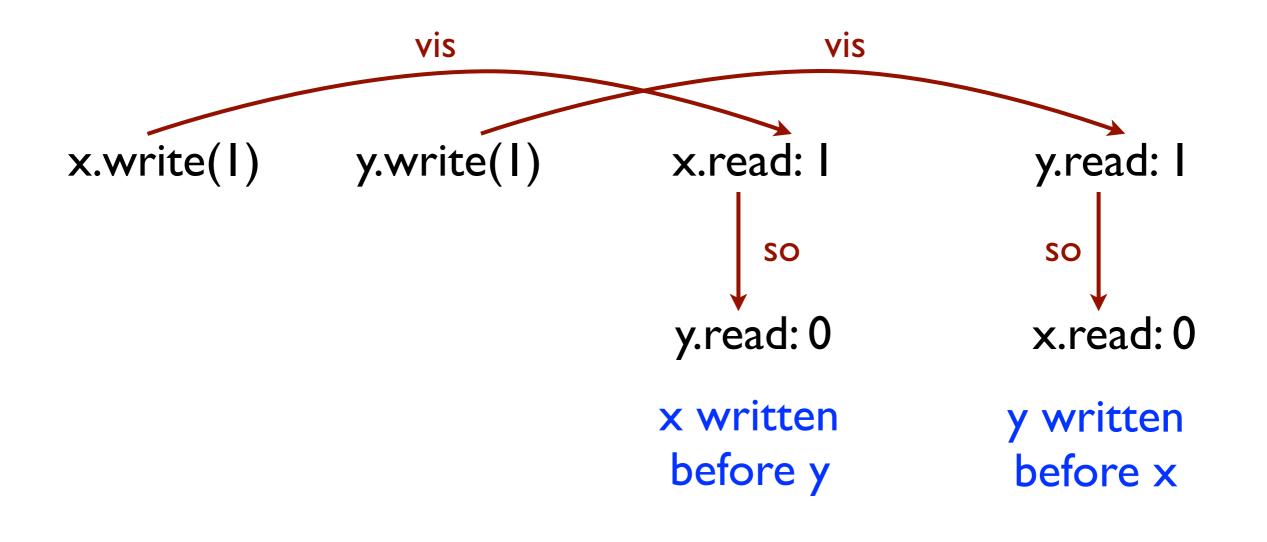




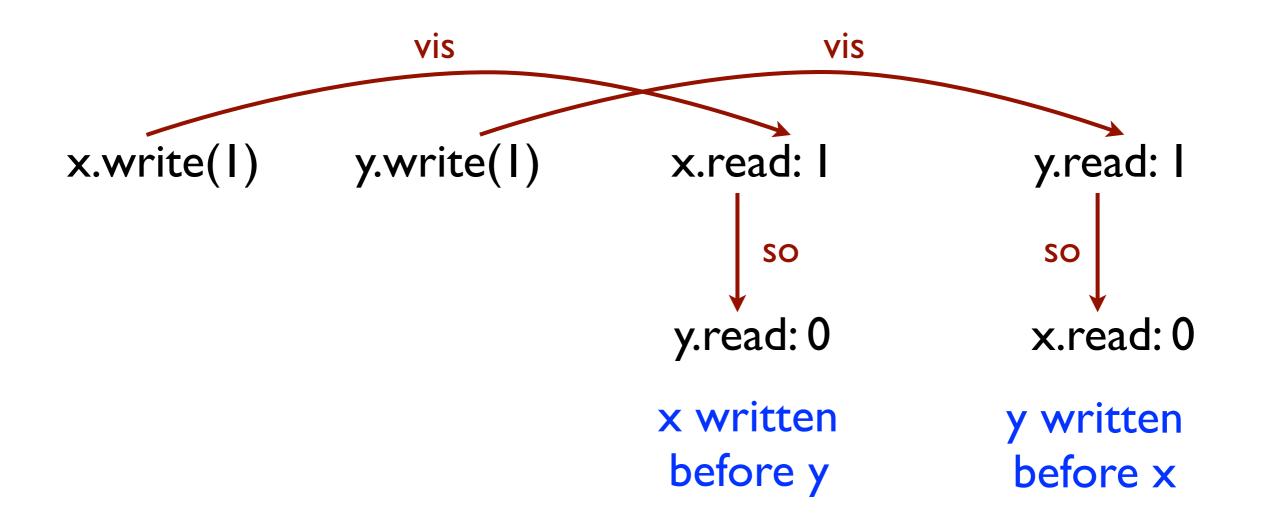
Independent reads of independent writes (IRIW)



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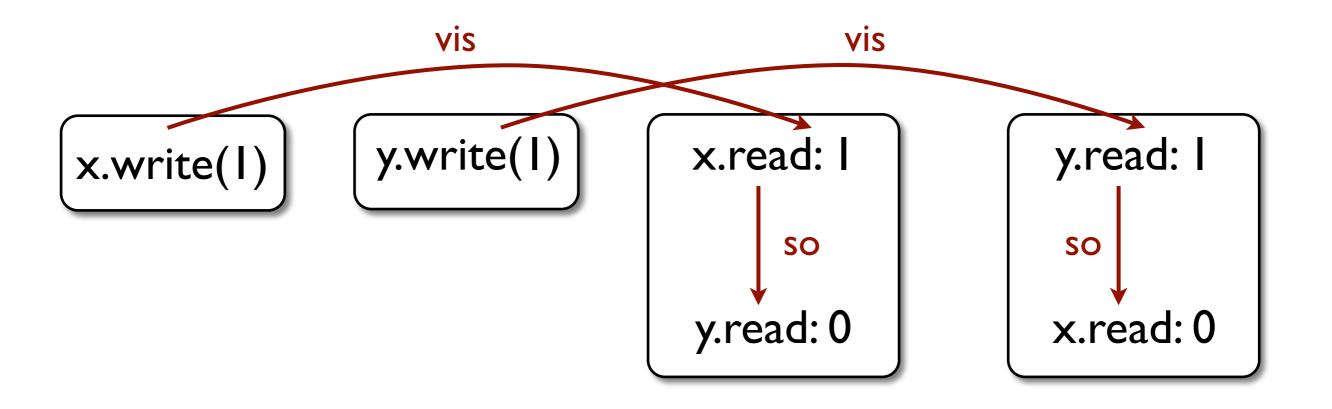


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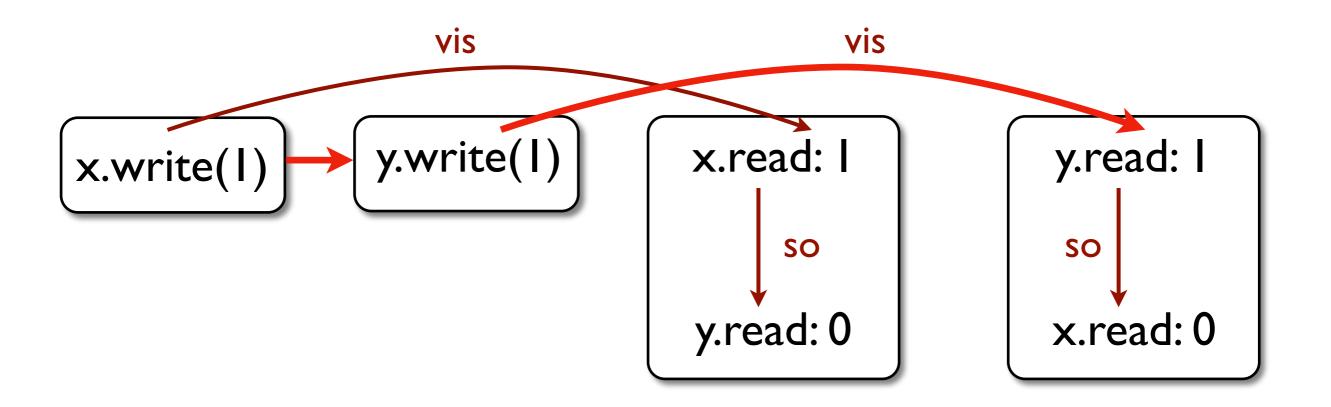


Implementations: no causal dependency between the two writes can be delivered in different orders at different replicas

Transactional IRIW = long fork



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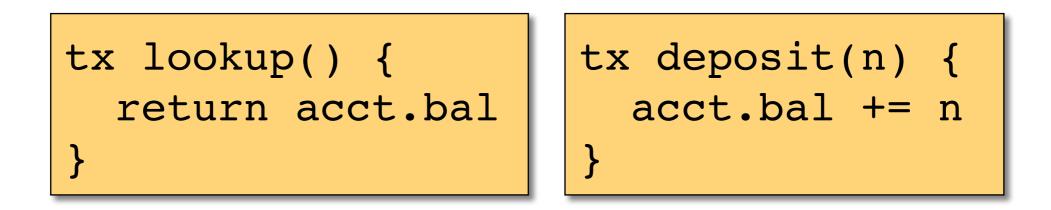
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Is an application robust against a particular consistency model?

Application behaves the same as when using a strongly consistent database

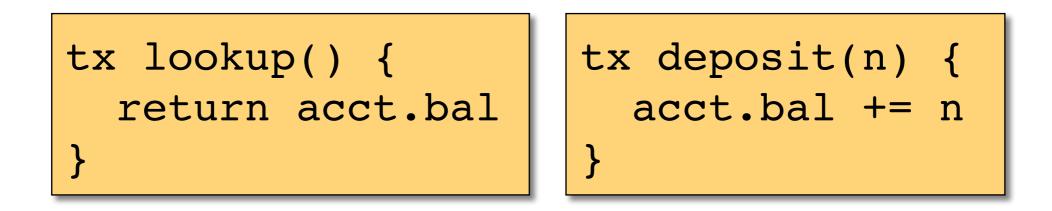
Application behaves the same whether using a PSI or a serializable database: [A]PSI = [A]SER

• Application: set of transactional programs {P₁, ..., P_n}



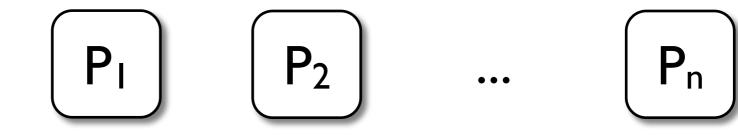
- Every program can generate multiple transactions at run time
- Simplification: every program is in its own session

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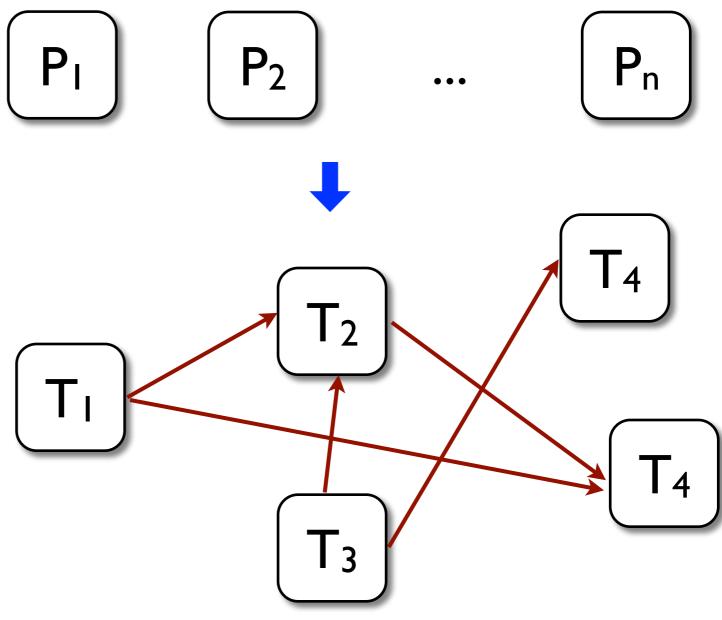
- Every program can generate multiple transactions at run time
- Simplification: every program is in its own session
- Checking robustness via static analysis: over-approximate the set of program behaviours

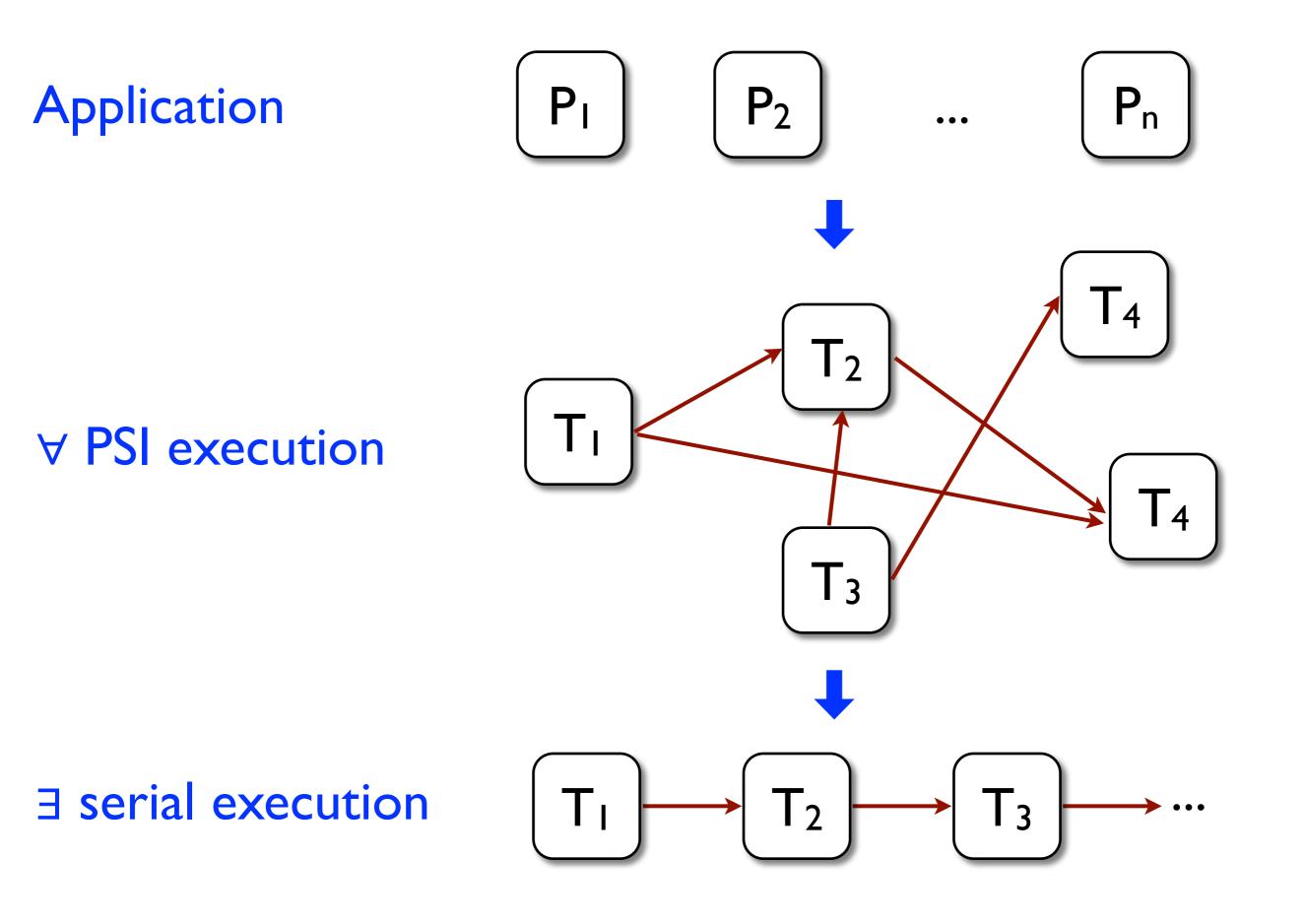


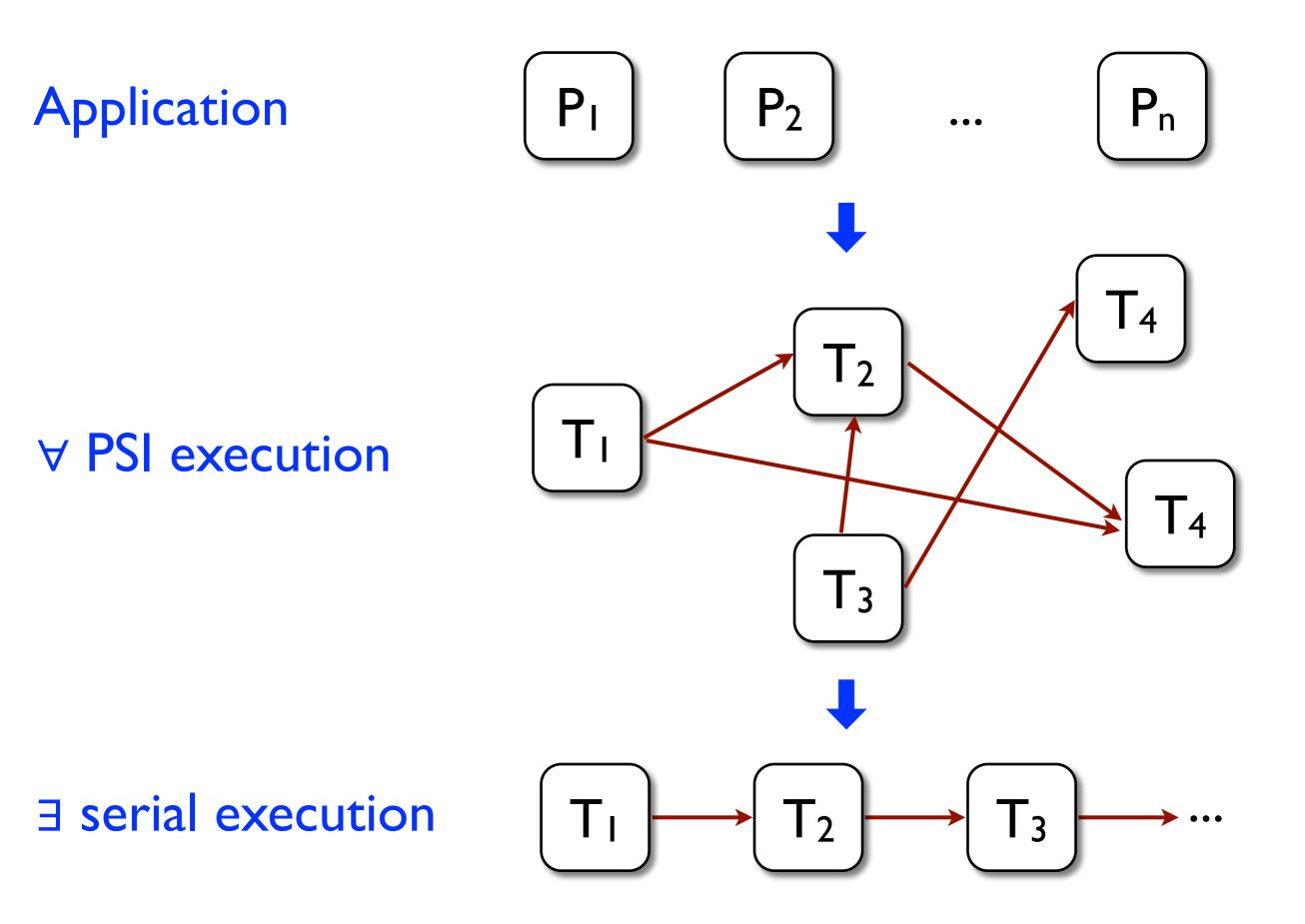




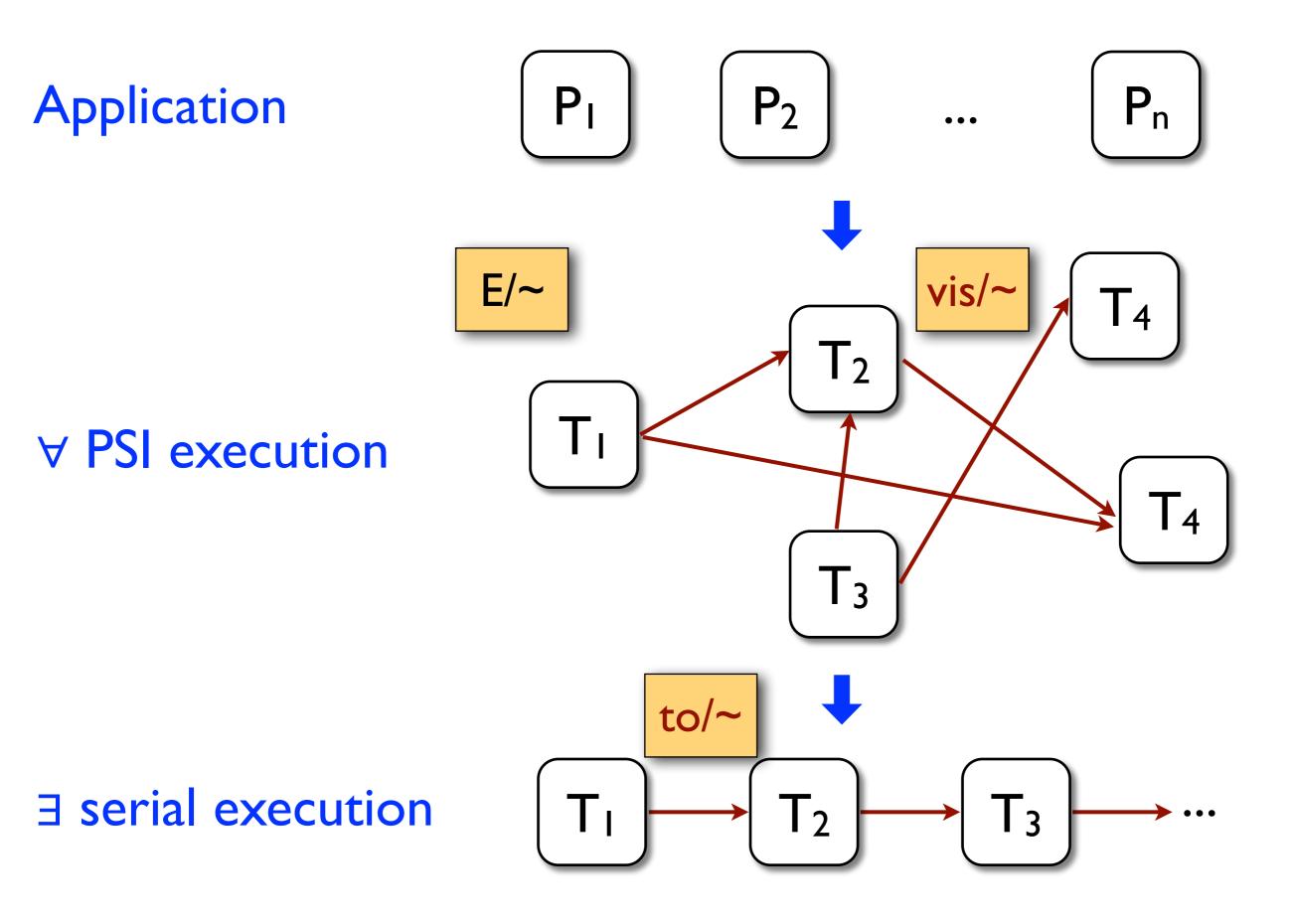
Application





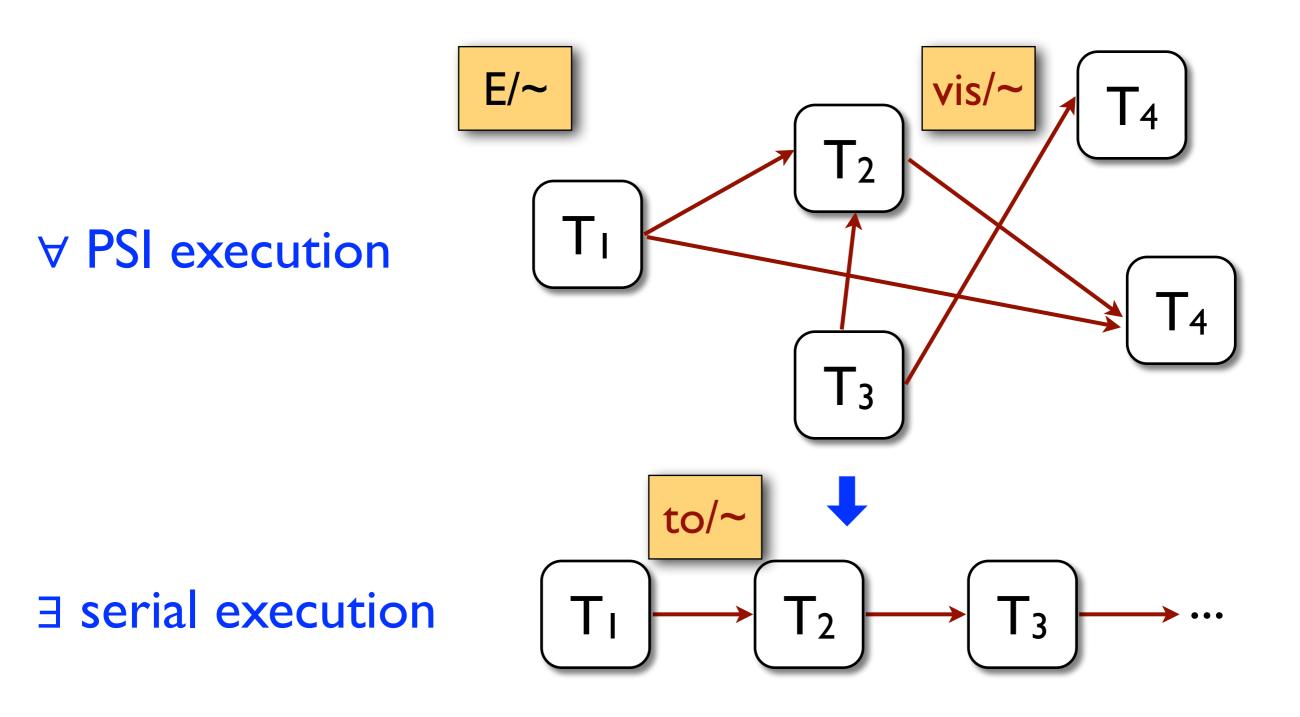


Each read returns the value written by the last write



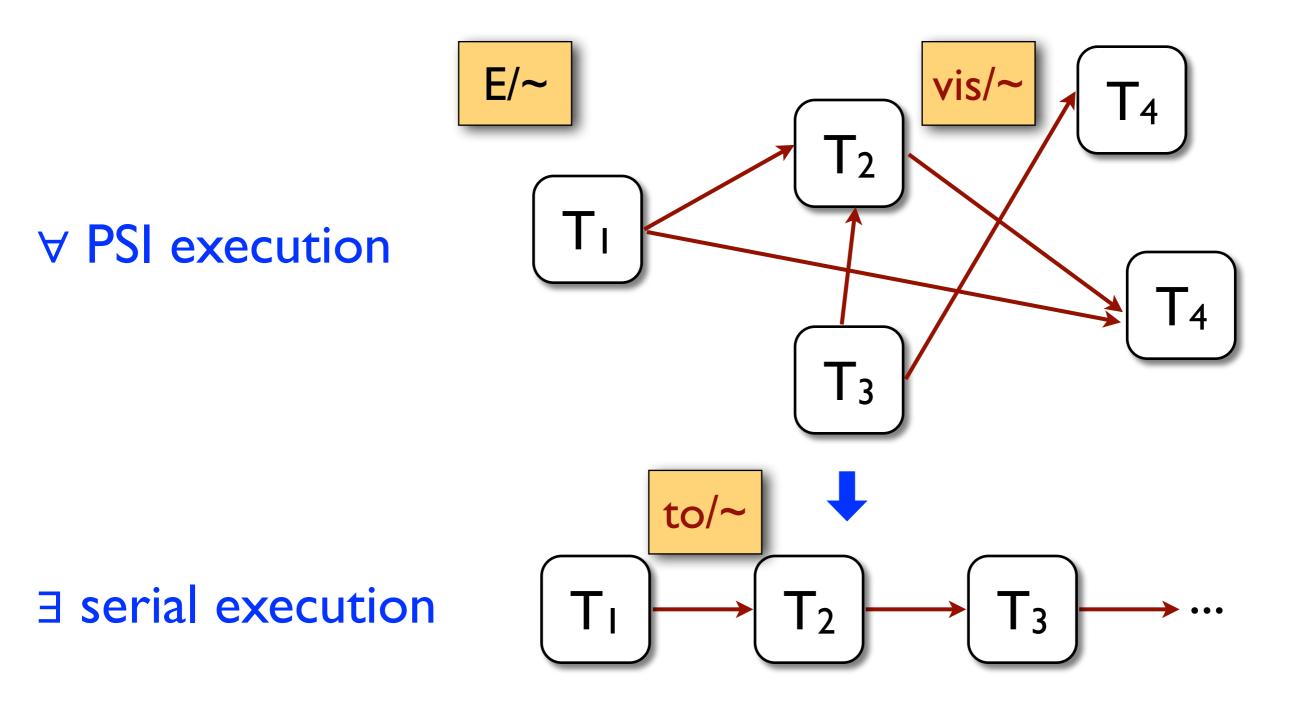
Each read returns the value written by the last write

First determine if a given PSI execution is serializable



Each read returns the value written by the last write

Build constraints on the serial order: relations on E/\sim that should be included into to/ \sim - transactional dependencies



Each read returns the value written by the last write

Write-read dependency (wr)

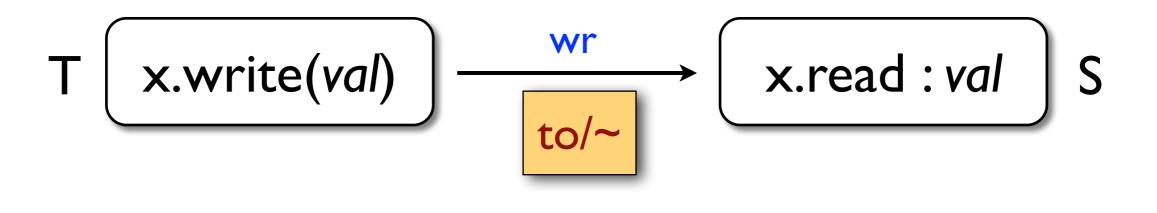
Given a PSI execution (E, ~, vis) and T, S \in E/~



 $T \xrightarrow{wr} S \iff S$ reads a value written by T

Write-read dependency (wr)

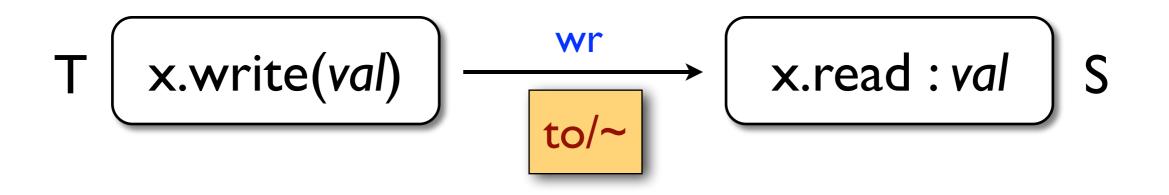
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Write-write dependency (wr)

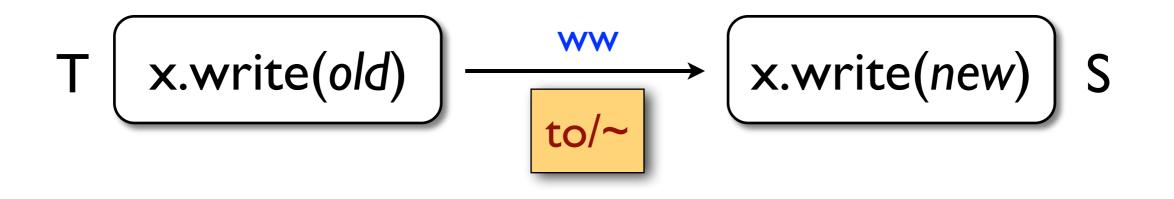
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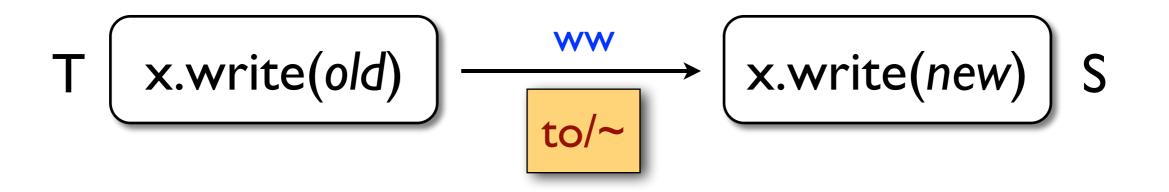
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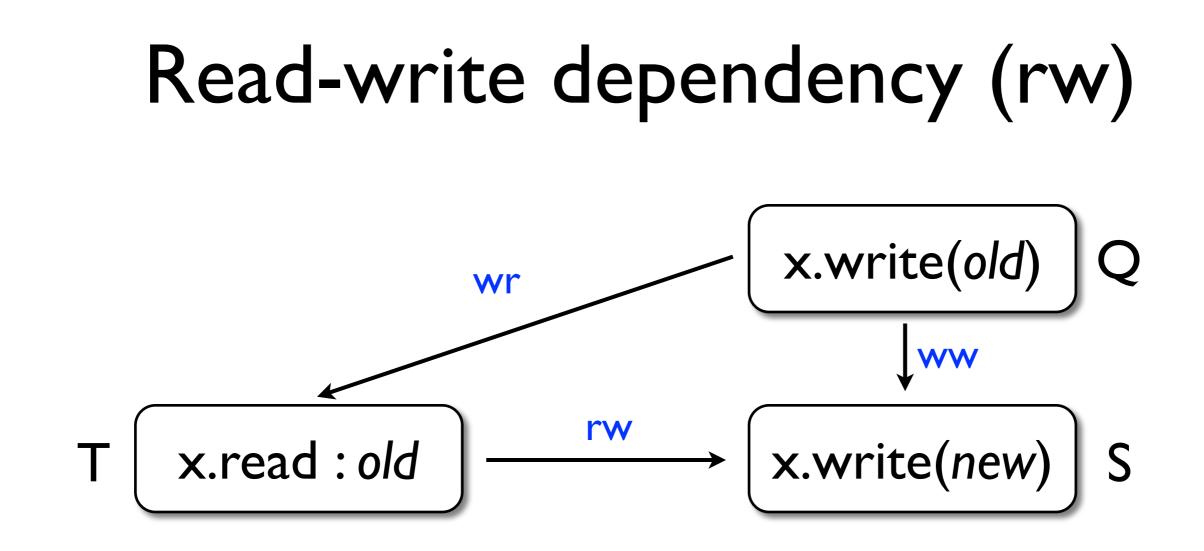


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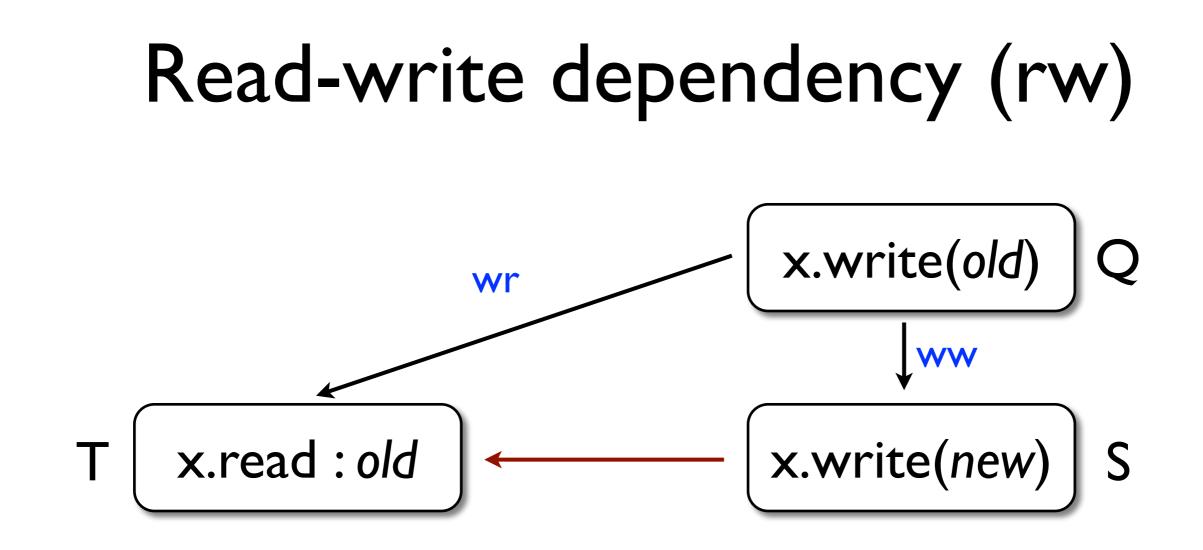
 $T \xrightarrow{ww}{\longrightarrow} S \iff T \text{ and } S \text{ contain writes to the same}$ object x and T $\xrightarrow{vis/\sim} S$

Read-write dependency (rw) T x.read : old \xrightarrow{rw} x.write(new) S

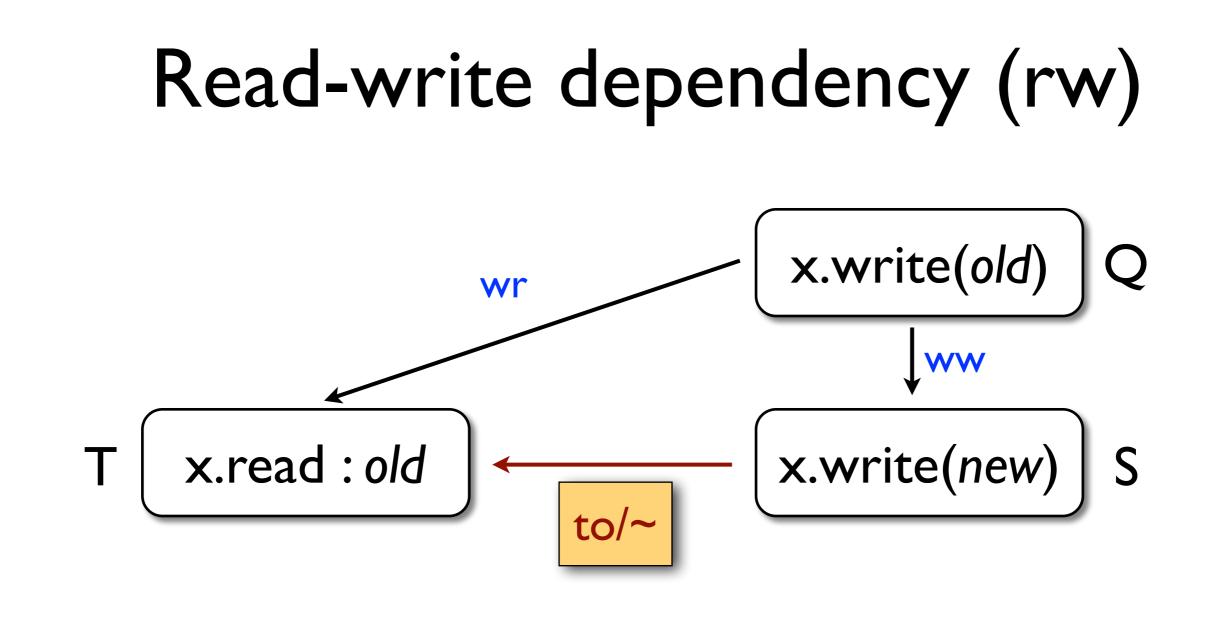
$$\mathsf{T} \xrightarrow{\mathsf{rw}} \mathsf{S} \Longleftrightarrow \mathsf{T} \neq \mathsf{S} \land \exists \mathsf{Q}. \mathsf{Q} \xrightarrow{\mathsf{wr}} \mathsf{T} \land \mathsf{Q} \xrightarrow{\mathsf{ww}} \mathsf{S}$$



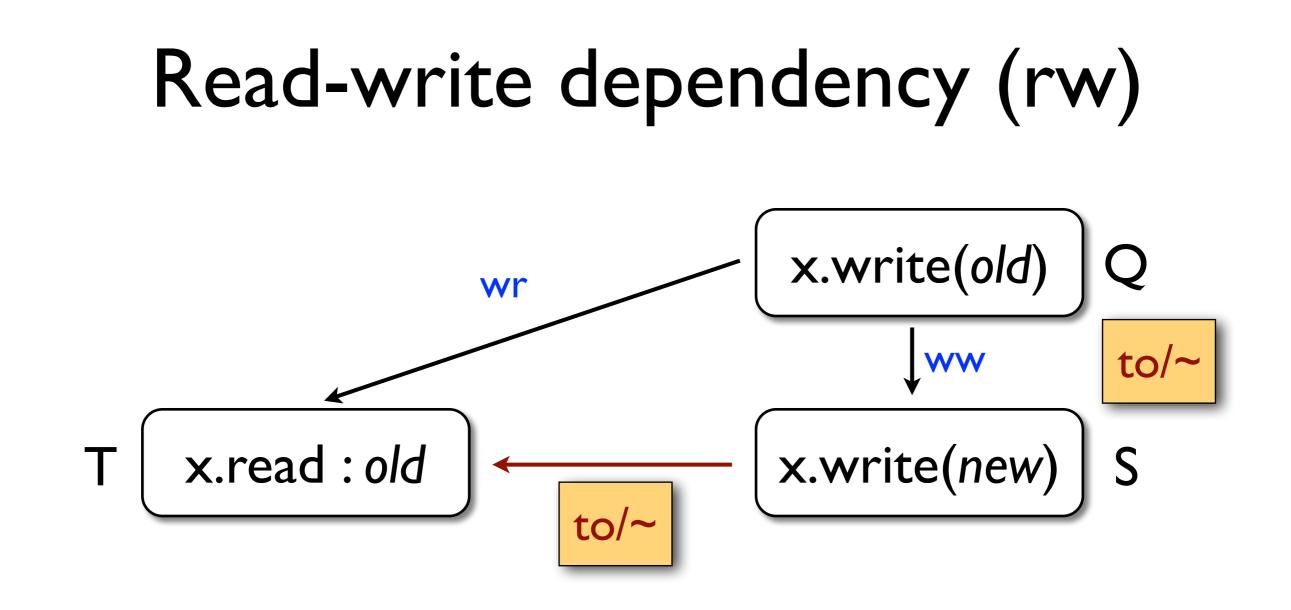
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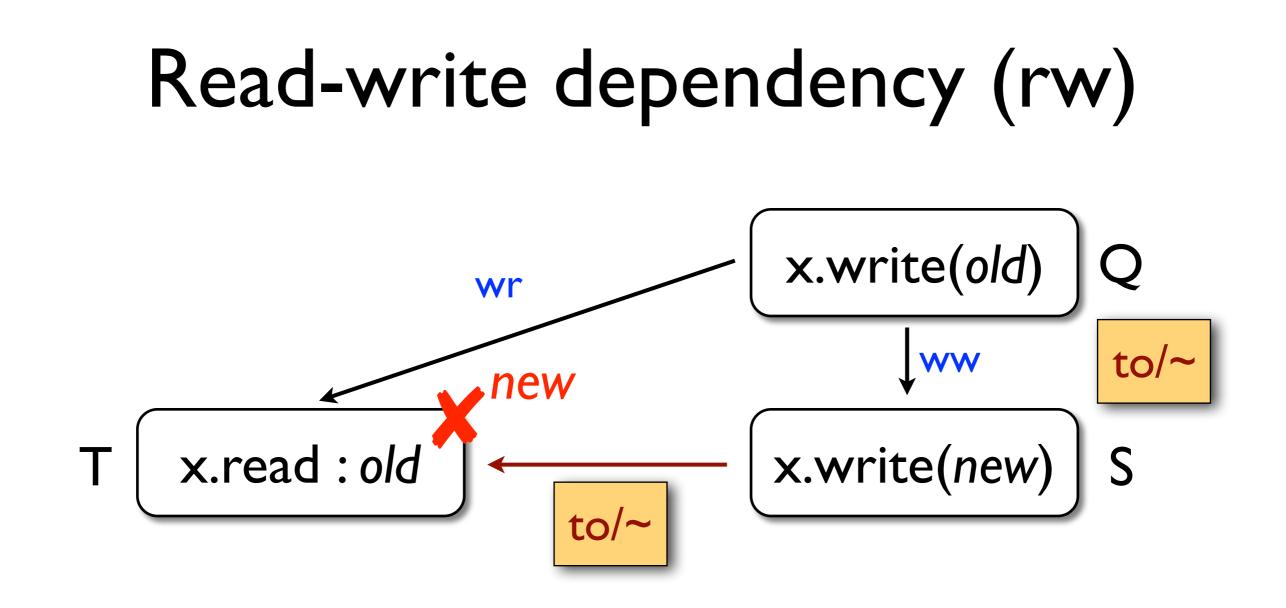
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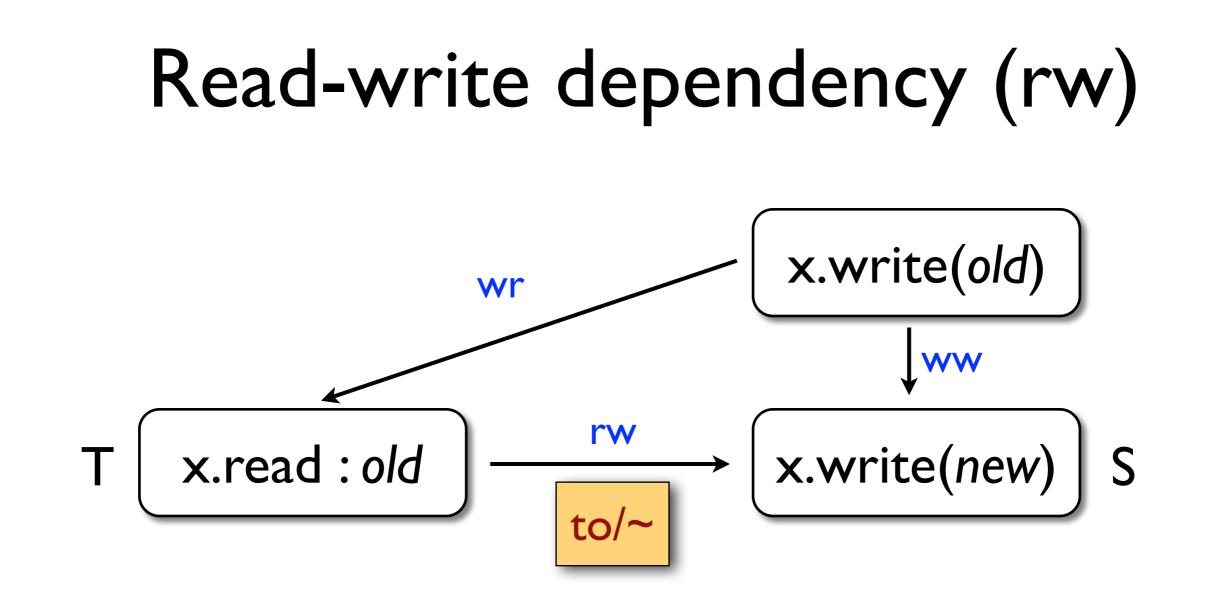
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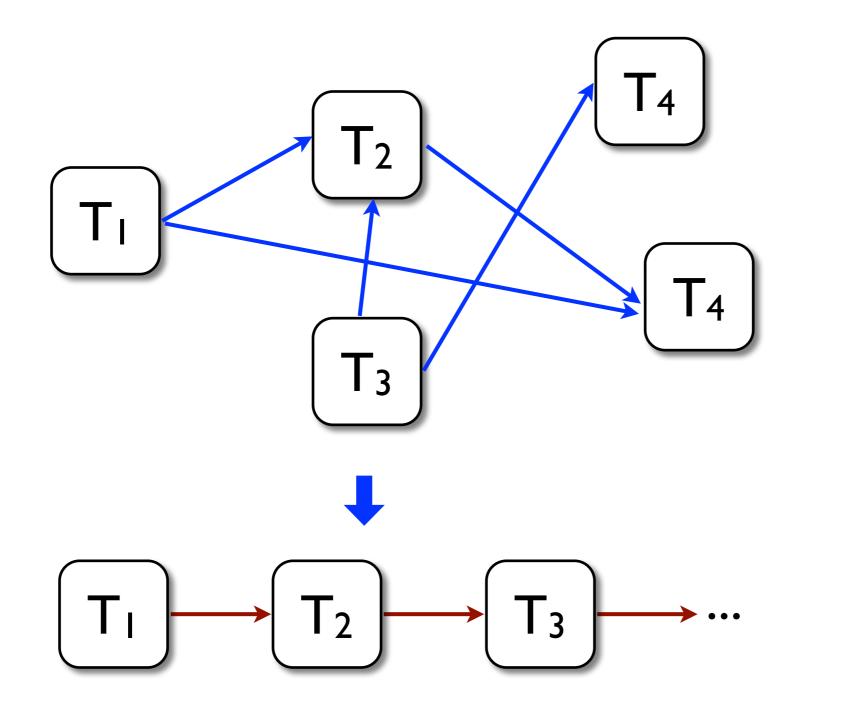


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Dependency graphs

- PSI execution (E, ~, vis) →
 dependency graph (E/~, wr, ww, rw)
- Theorem: If the dependency graph is acyclic, then the execution is serializable

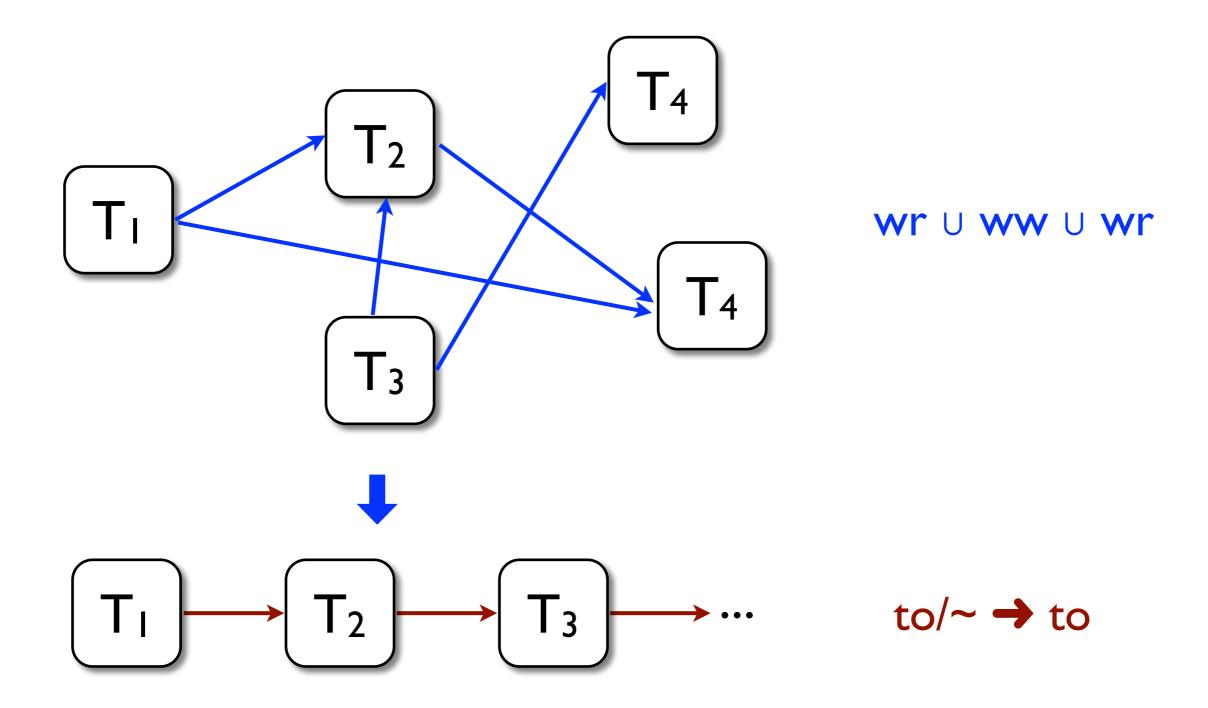
If (wr \cup ww \cup wr) is acyclic, then there is a total order on E/~ containing it [order-extension principle] \rightarrow the desired order to



wr U ww U wr



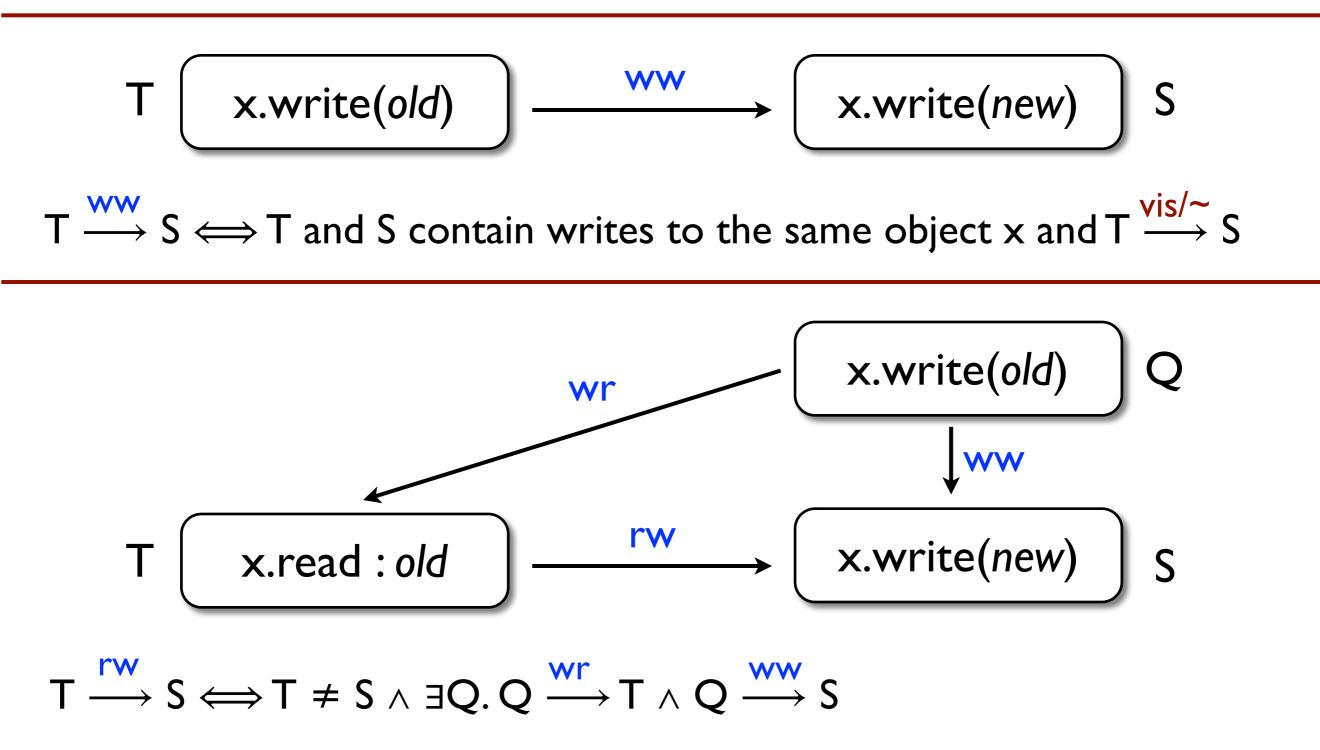
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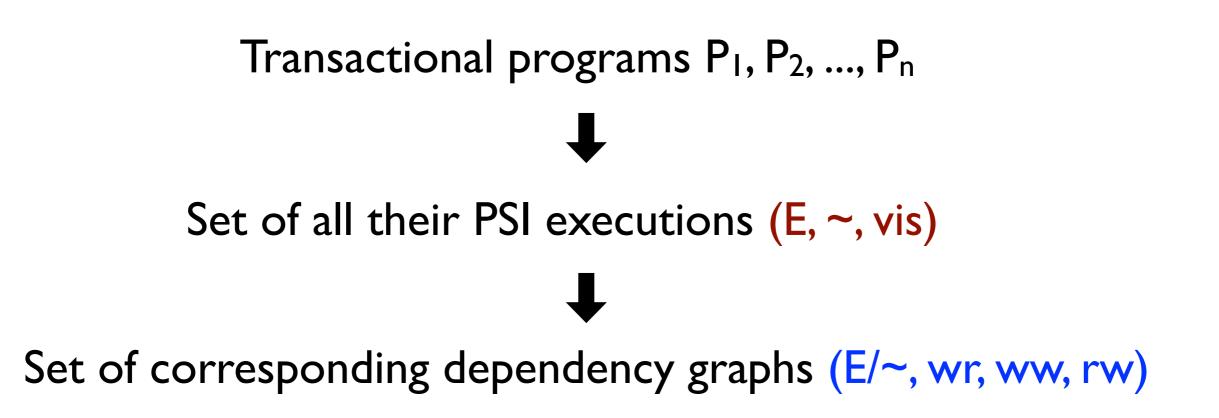
Each read returns the value written by the last write in to?

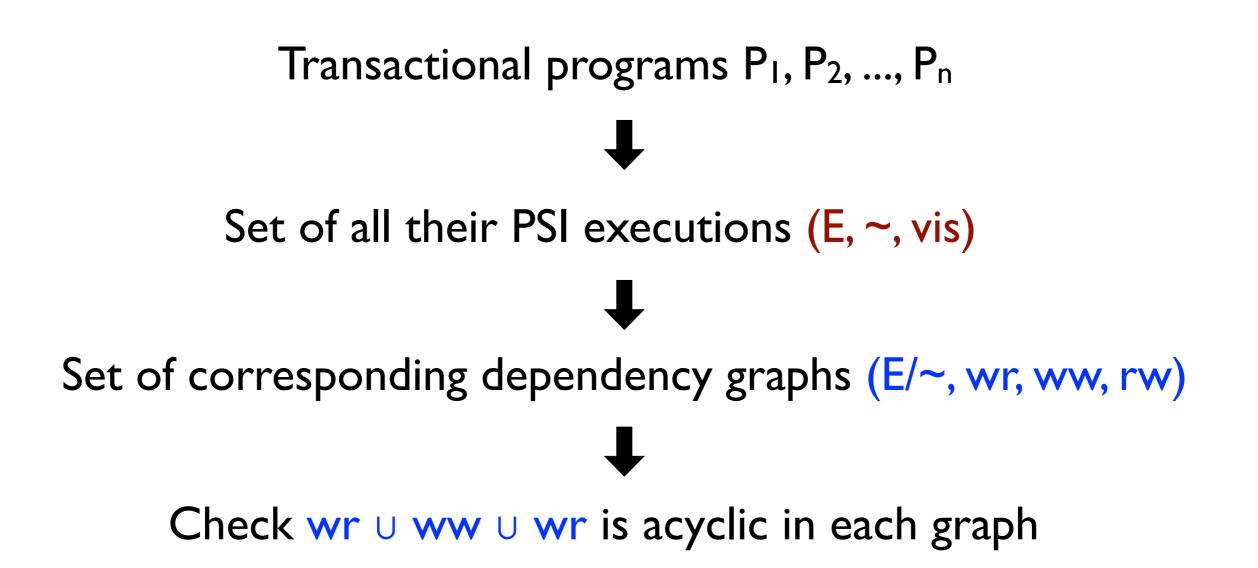


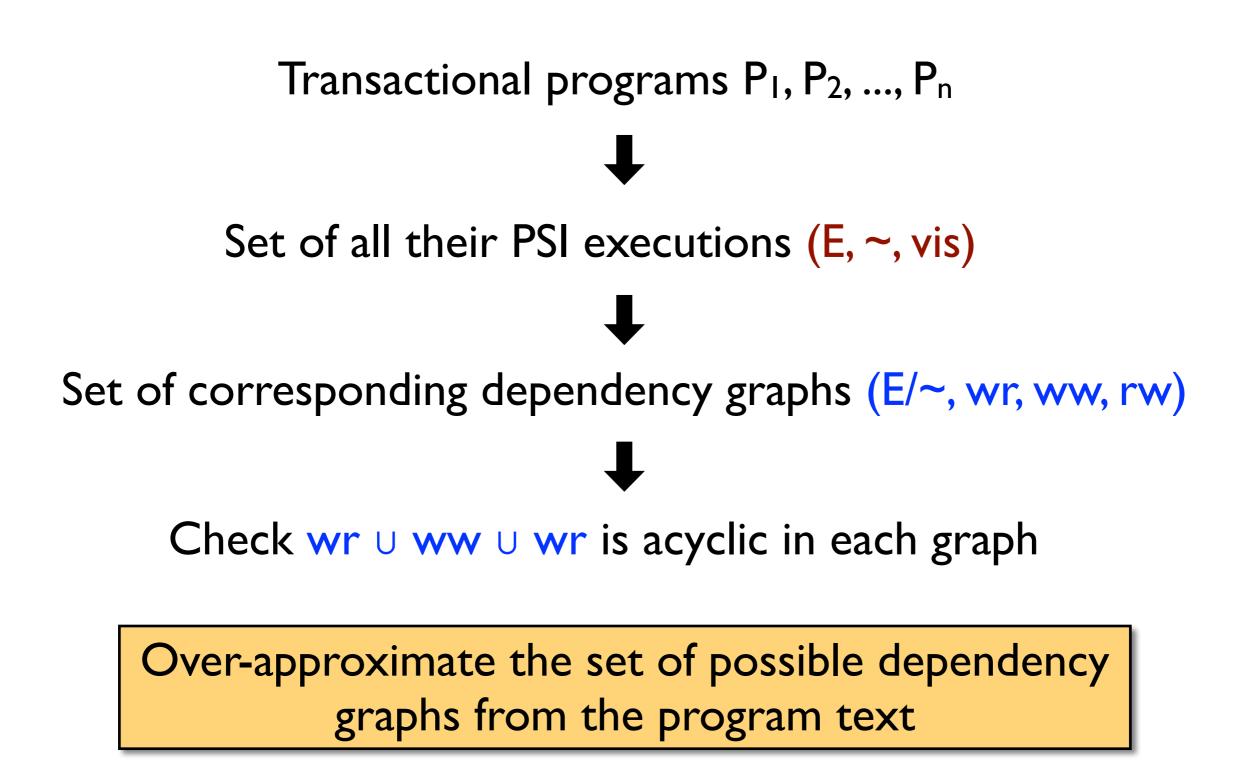
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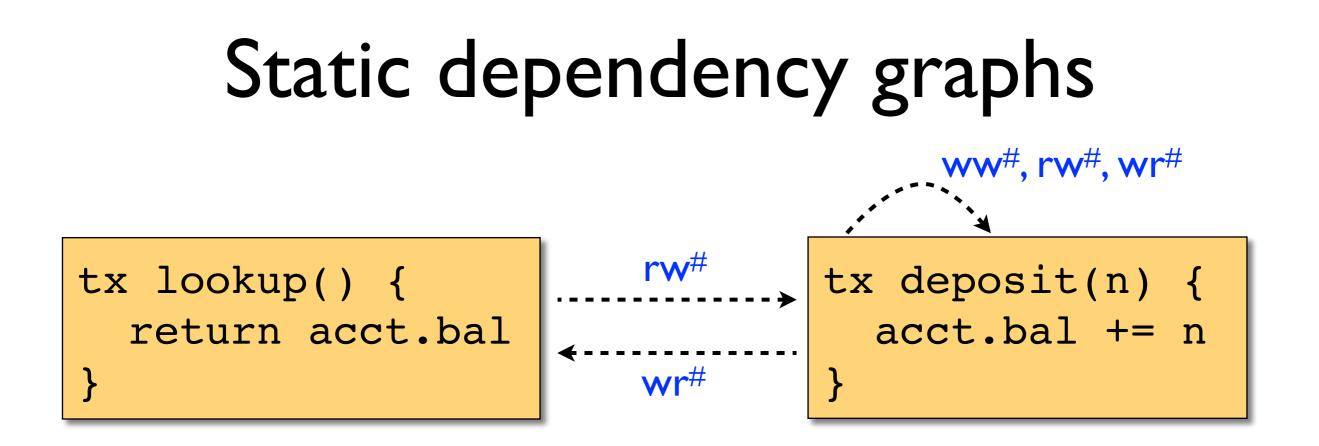


Transactional programs P₁, P₂, ..., P_n Set of all their PSI executions (E, ~, vis)

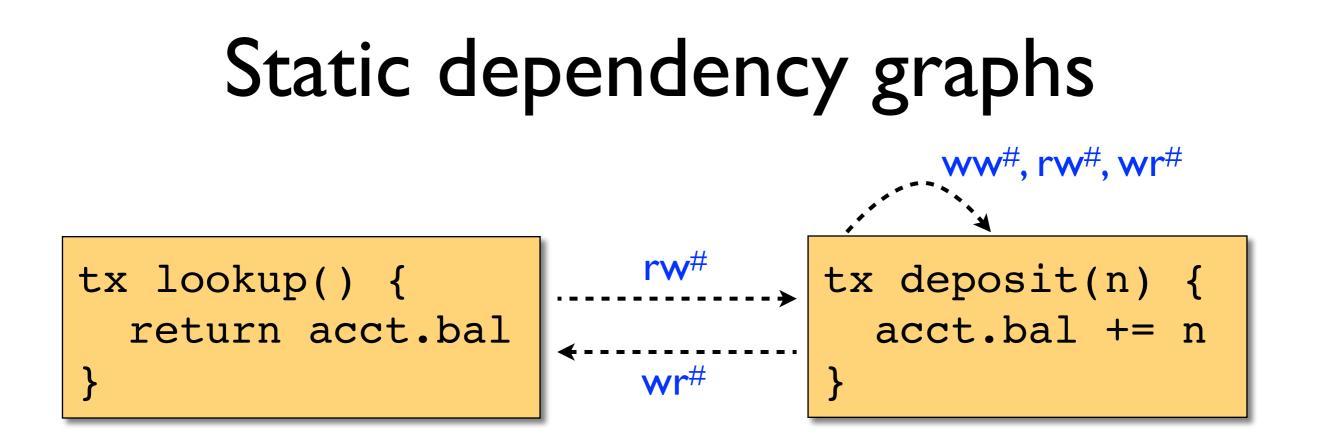




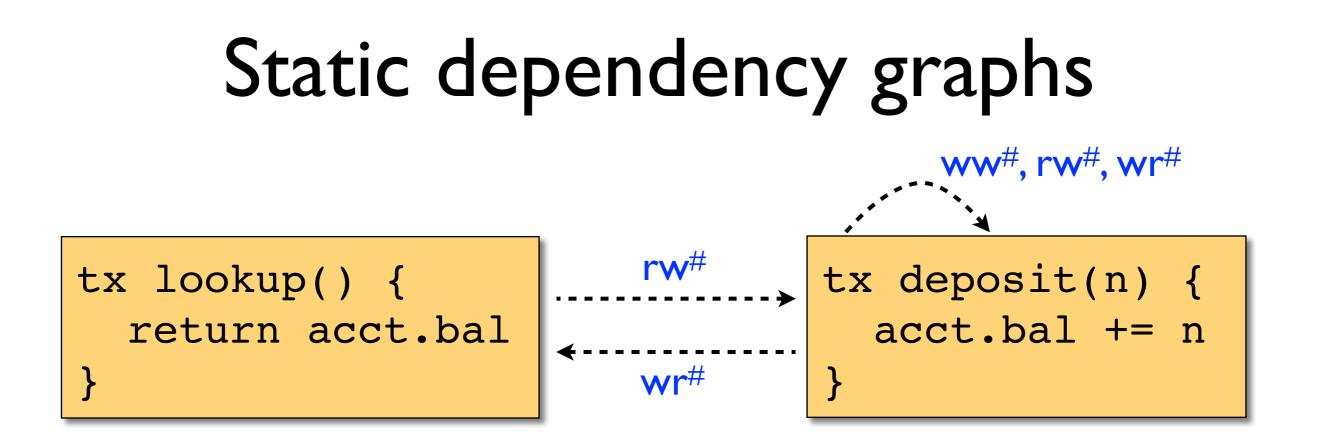




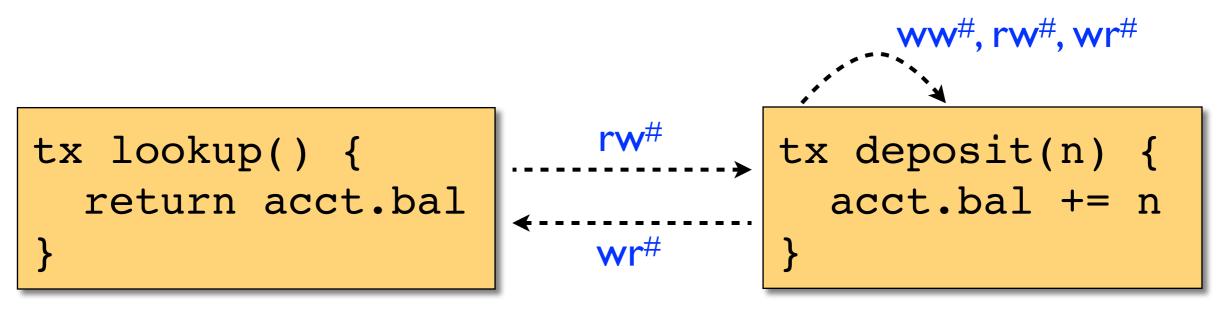
- Nodes: transactional programs
- Edges: over-approximations of dependencies wr[#], ww[#], rw[#]

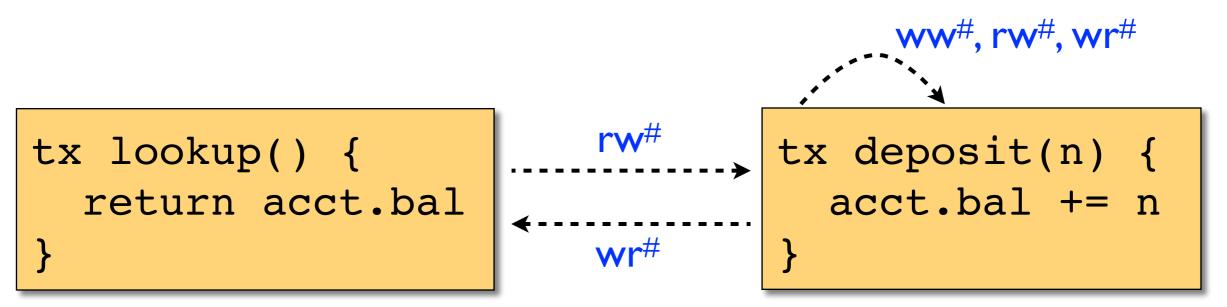


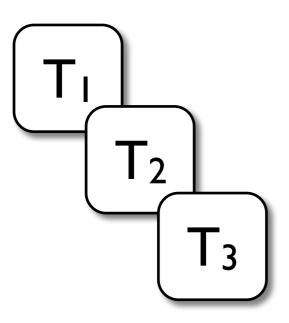
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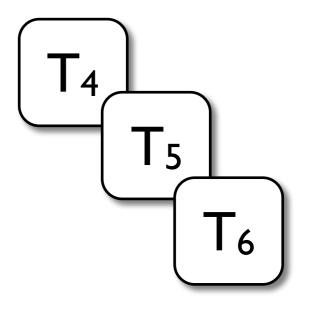


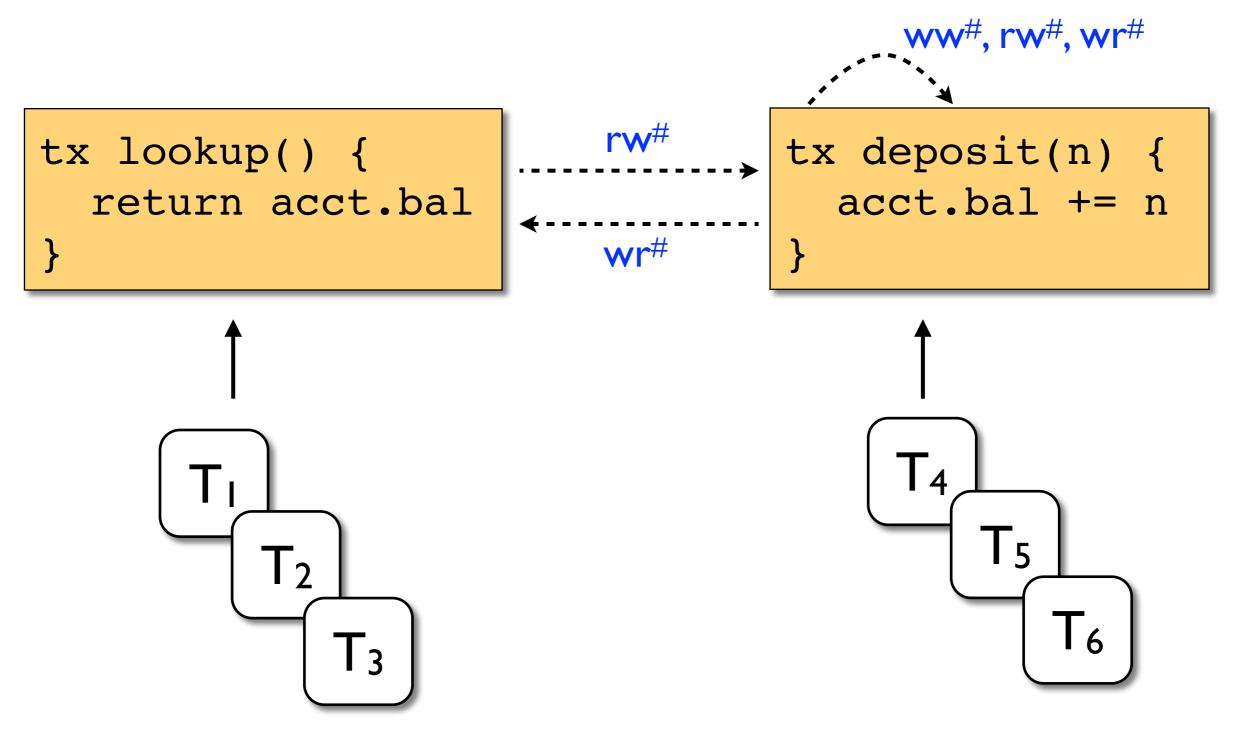
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- Represents an over-approximation of all dynamic dependency graphs that can be produced by the programs



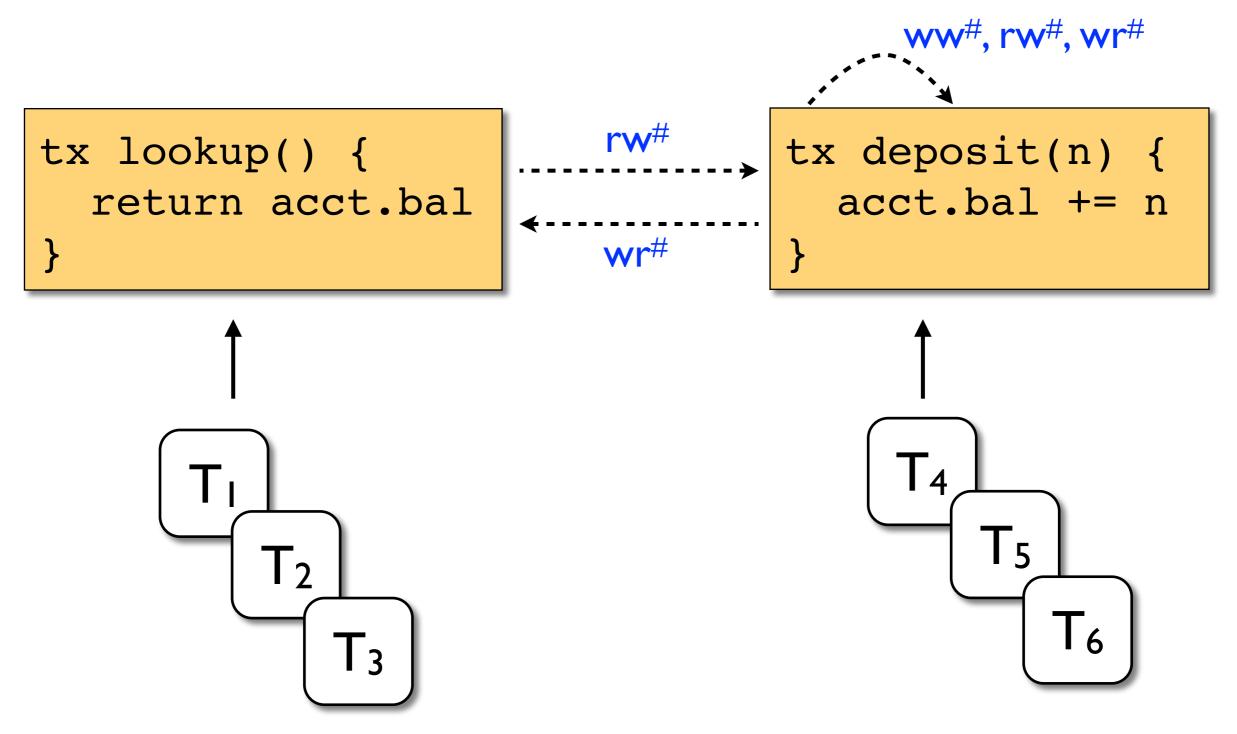


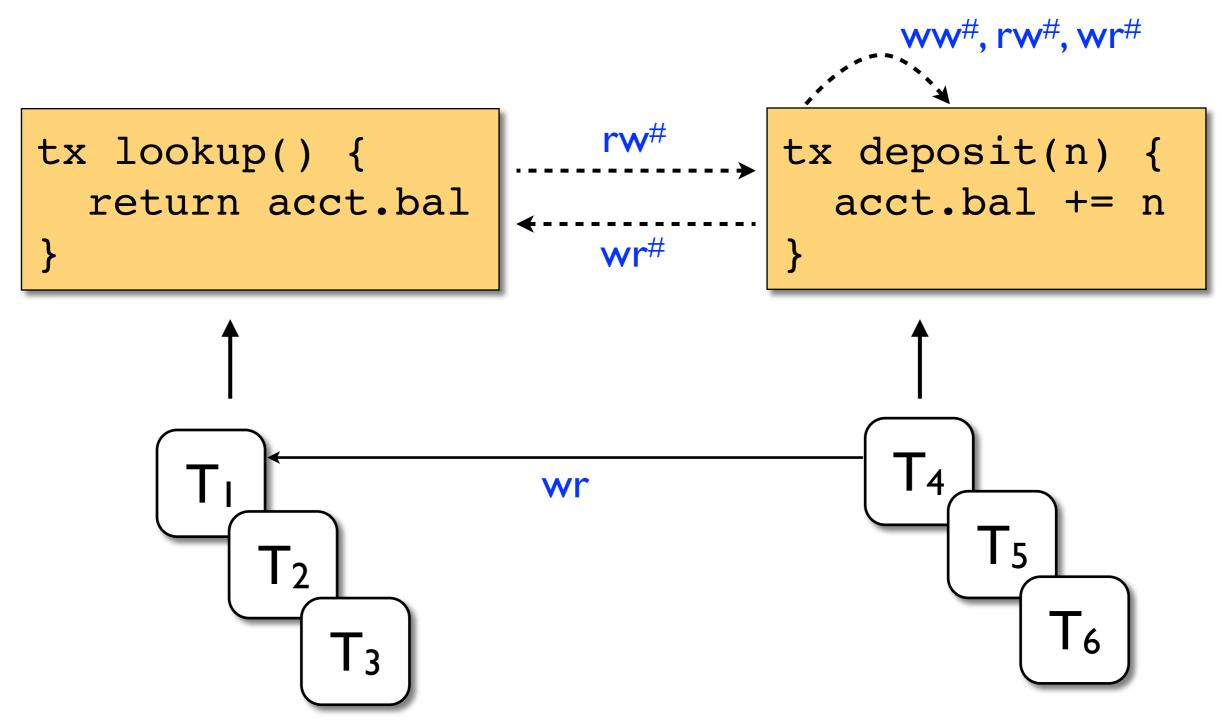


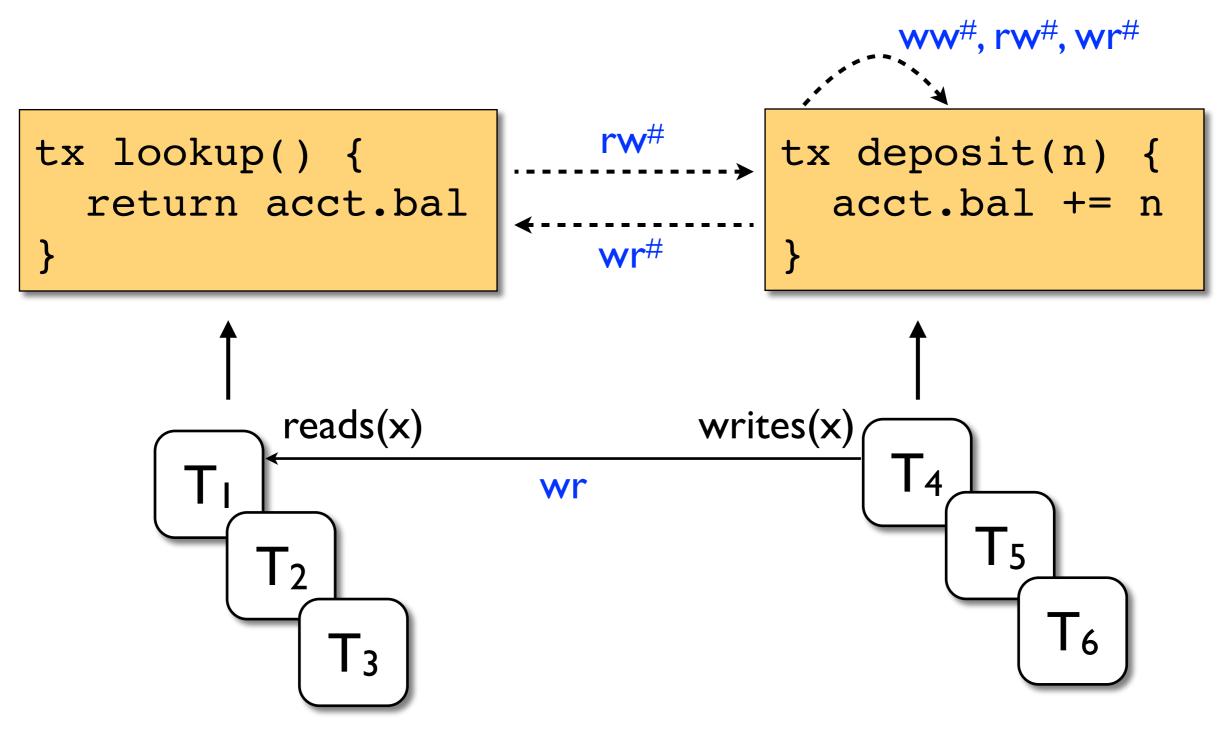


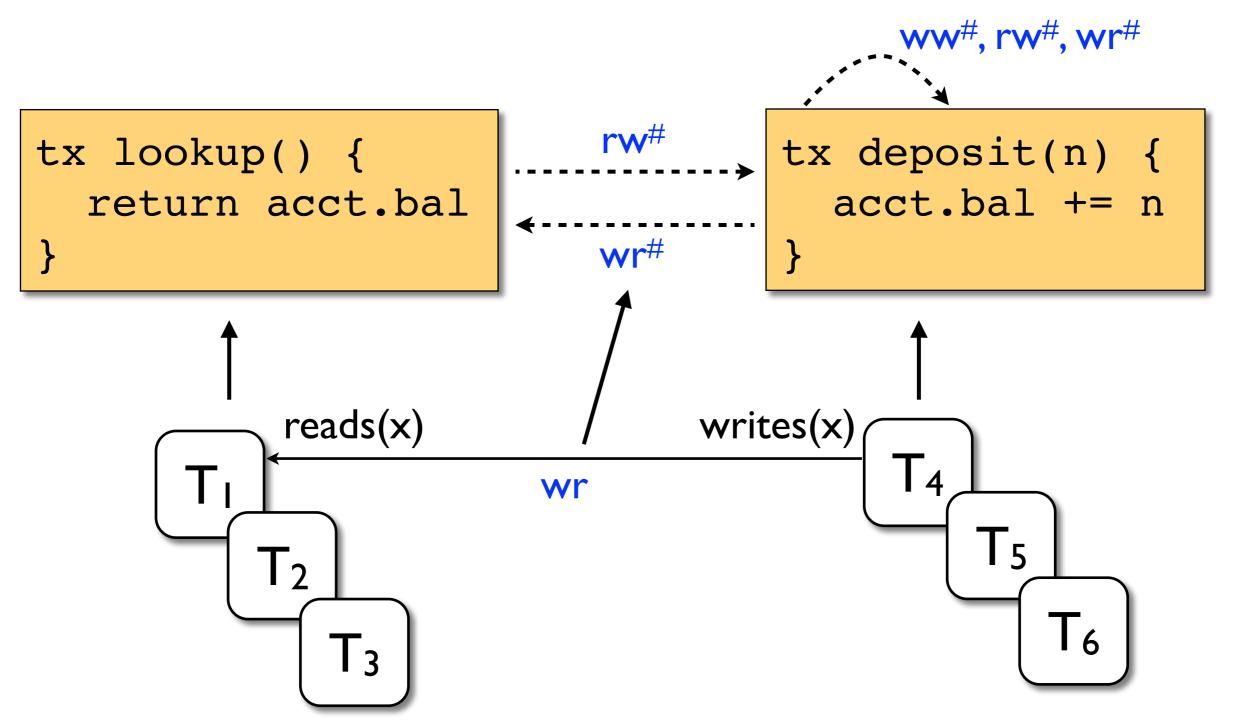


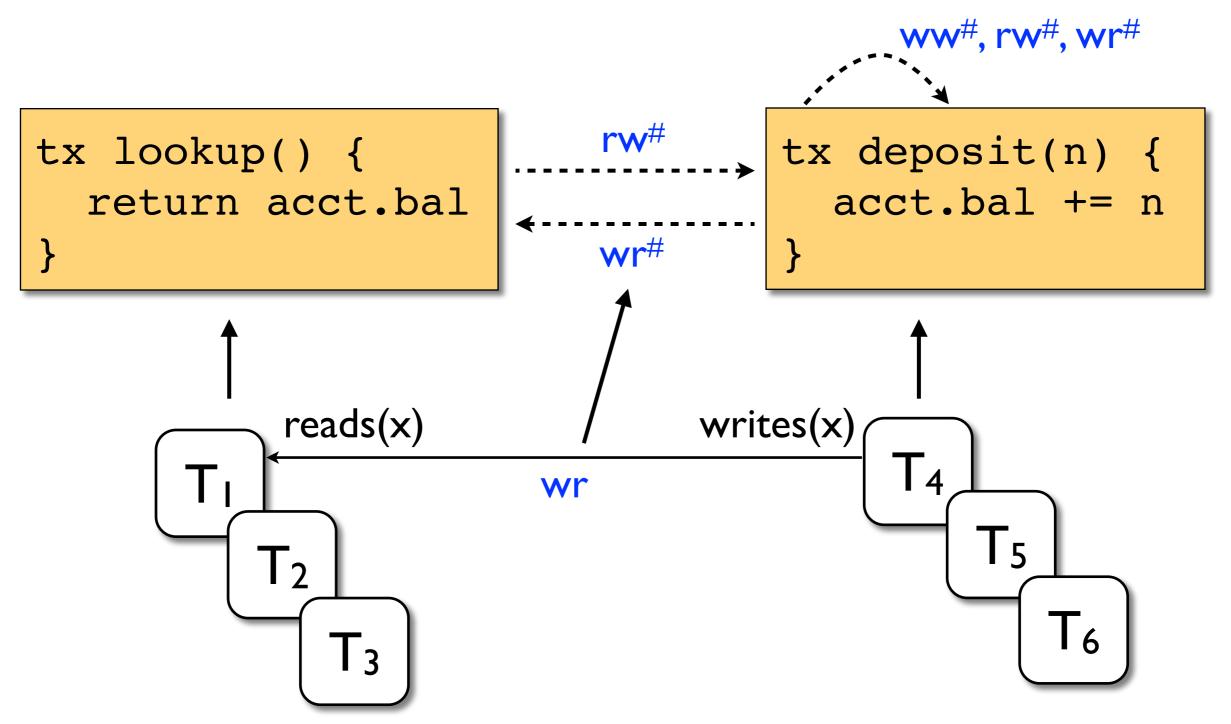
Transactions arising from the same program map to the same node



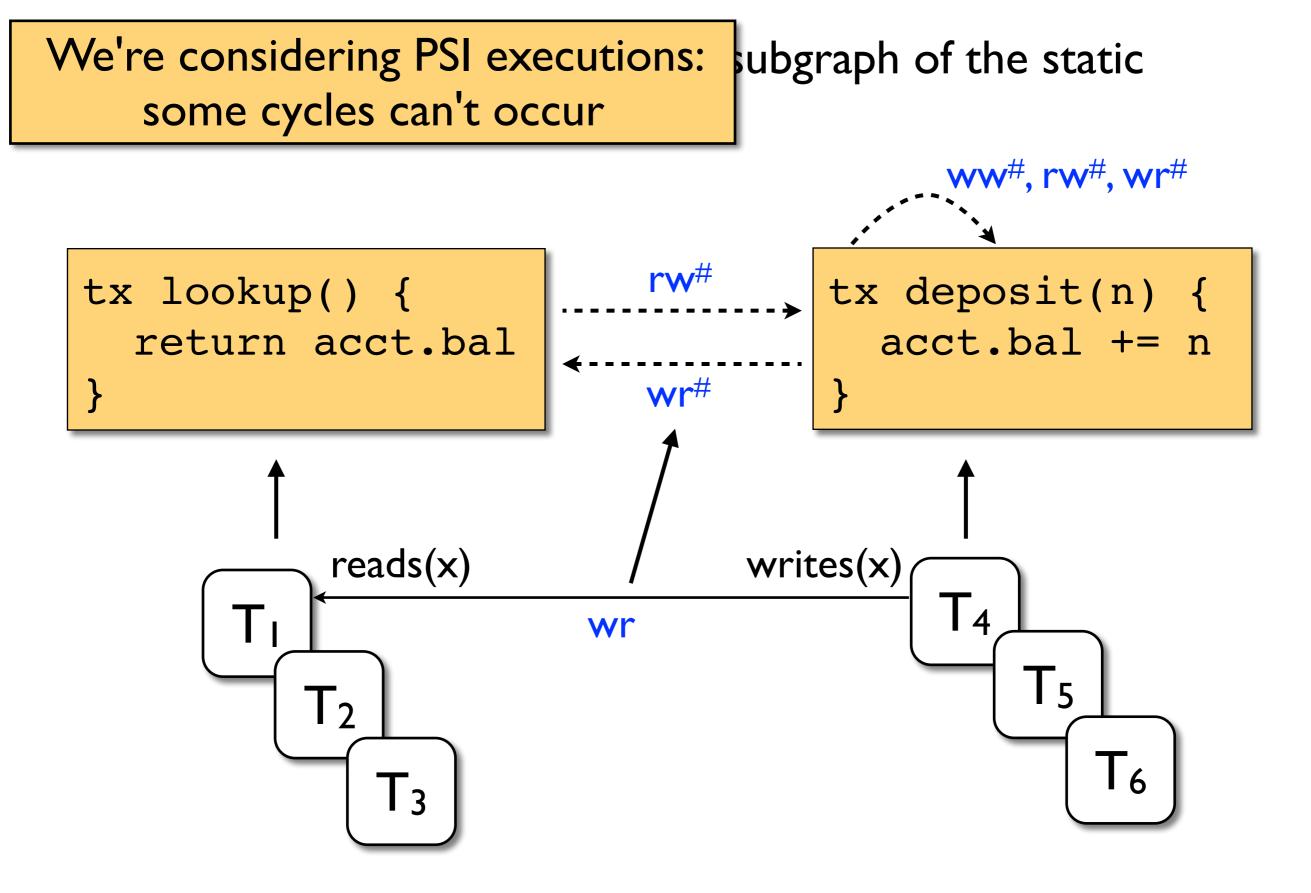








Cycle in the dynamic graph \rightarrow cycle in the static graph If the static graph is acyclic, so is the dynamic one



Cycle in the dynamic graph \rightarrow cycle in the static graph If the static graph is acyclic, so is the dynamic one



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$$T \left[x.write(old) \right] \xrightarrow{ww} \left[x.write(new) \right] S$$

 $T \xrightarrow{ww} S \iff T$ and S contain writes to the same object x and $T \xrightarrow{vis/\sim} S$



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wr \cup ww \subseteq vis/~ - acyclic

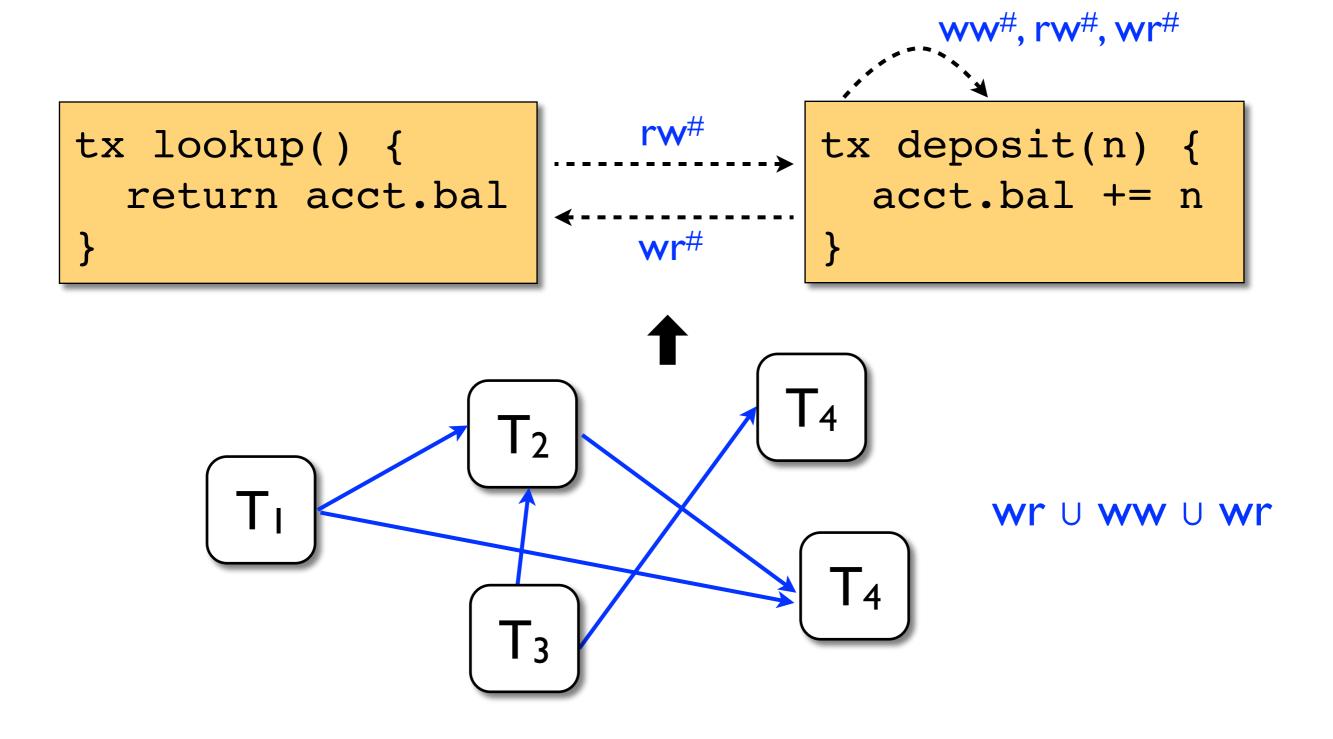


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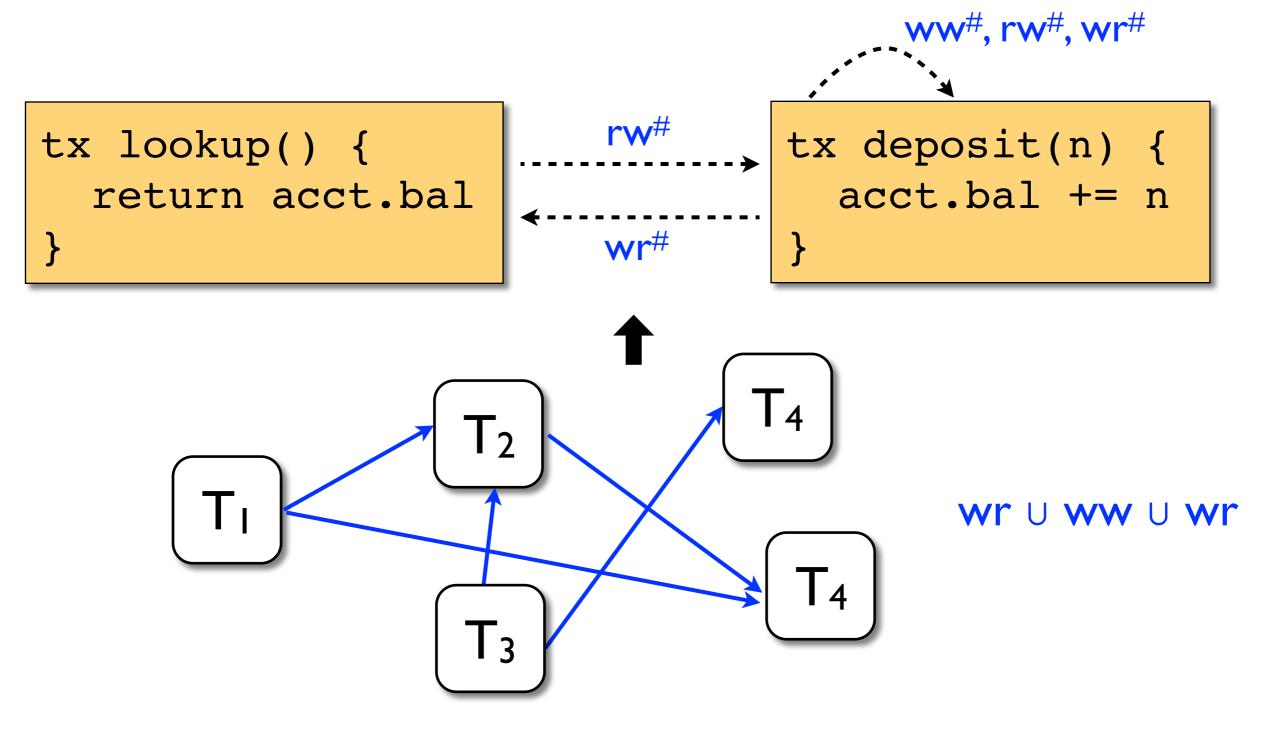
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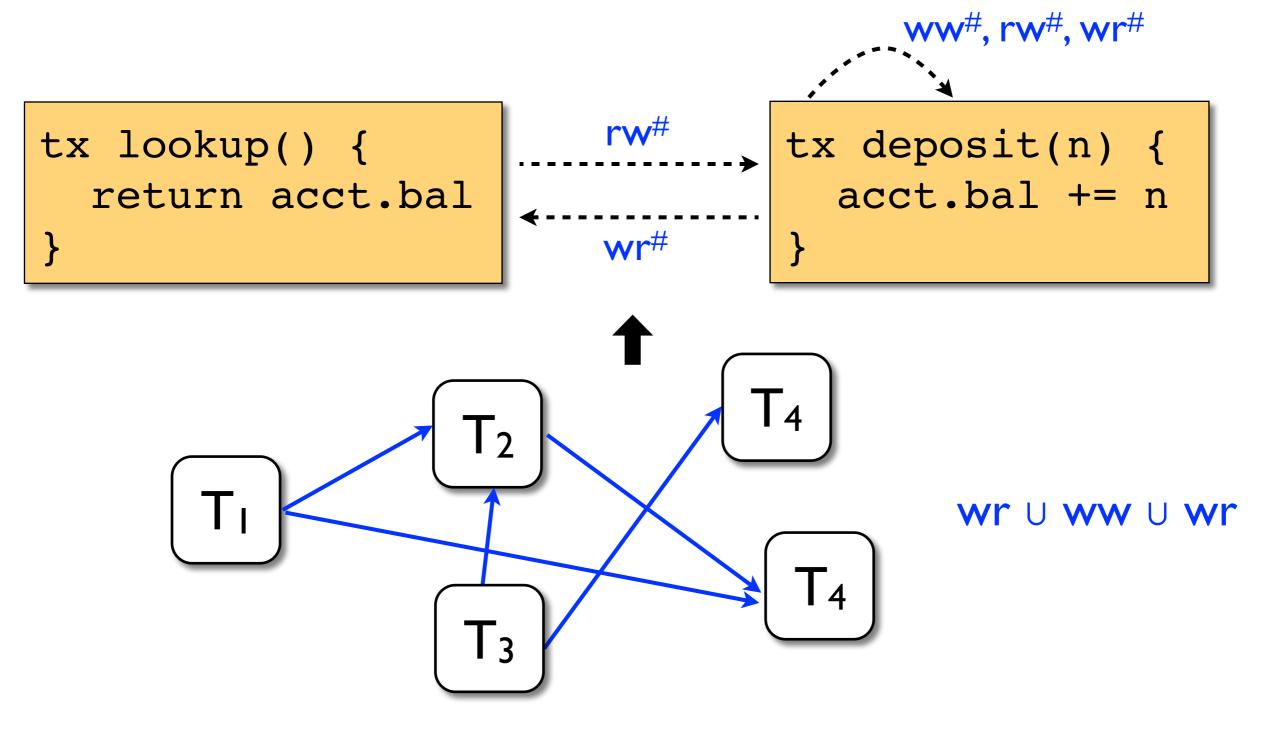
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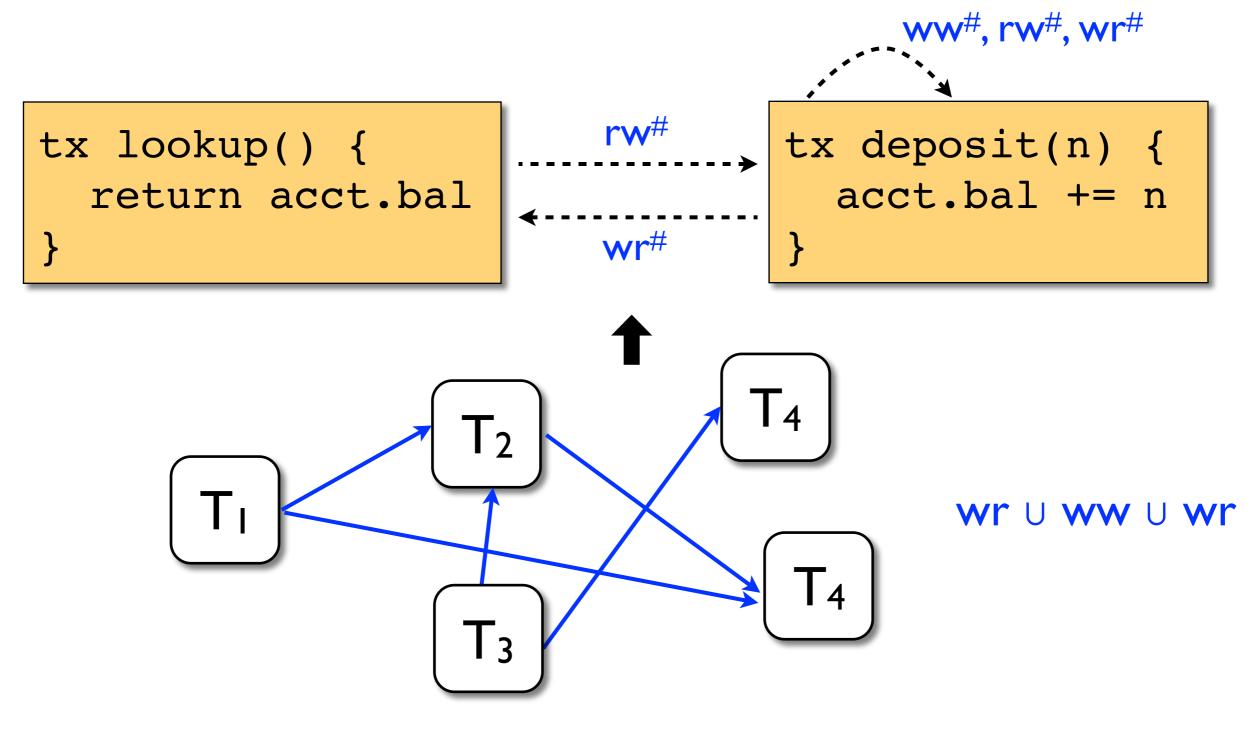
Dynamic dependency graph \rightarrow a subgraph of the static dependency graph



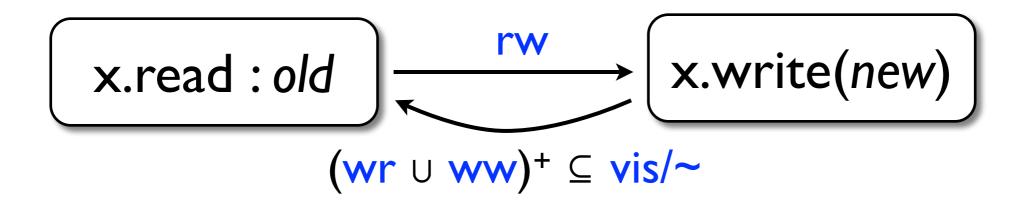
 Dynamic cycles with no rw edges aren't PSI → don't represent robustness violations

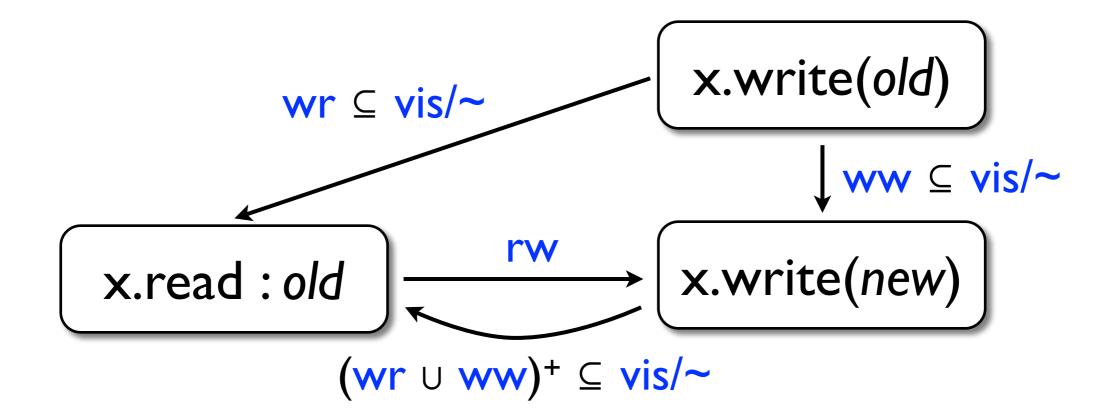


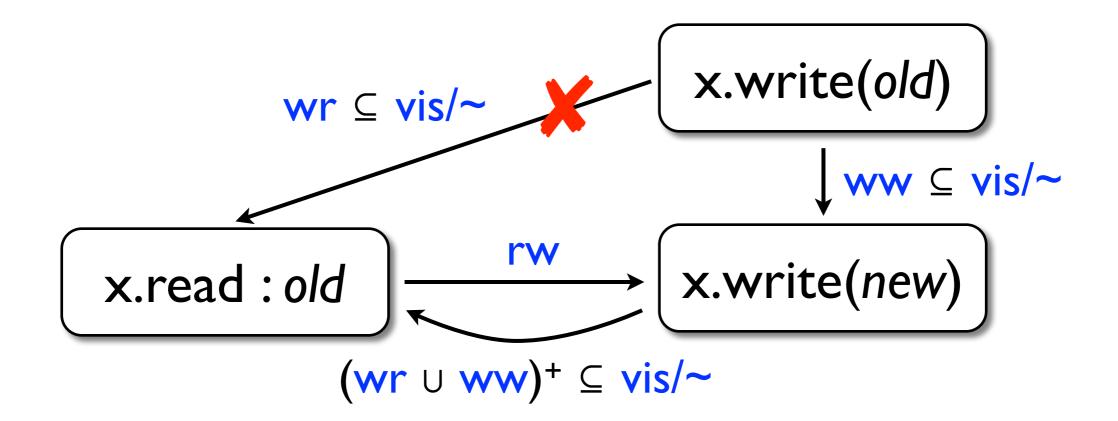
- Dynamic cycles with no rw edges aren't PSI → don't represent robustness violations
- Enough to check no cycles in (wr \cup ww \cup rw) with \geq I rw



- Dynamic cycles with no rw edges aren't PSI → don't represent robustness violations
- Enough to check no cycles in (wr \cup ww \cup rw) with \geq I rw
- Enough to check no cycles in (wr[#] \cup ww[#] \cup rw[#]) with \geq I rw[#]





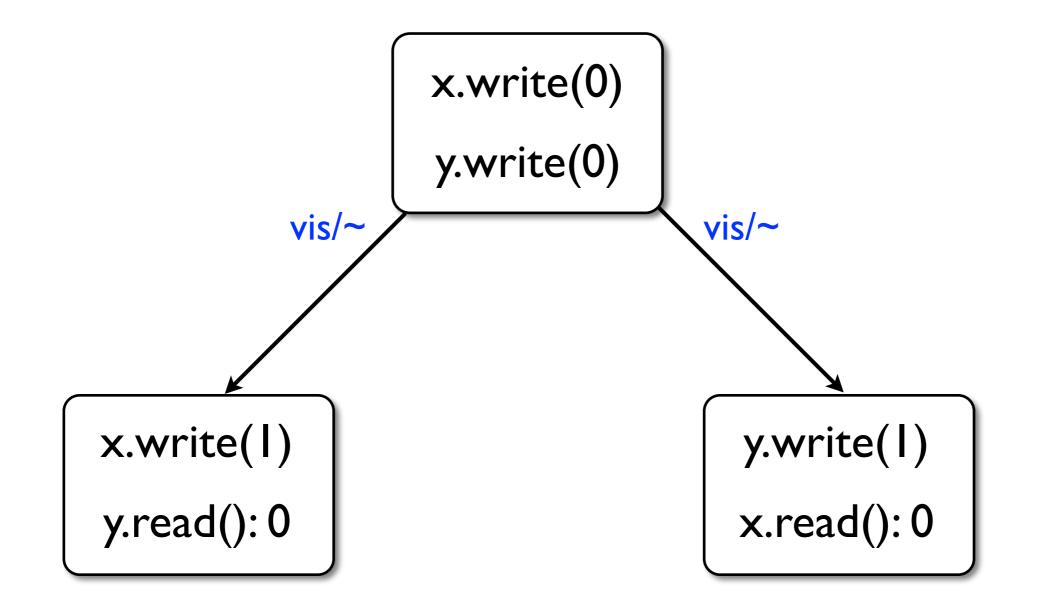


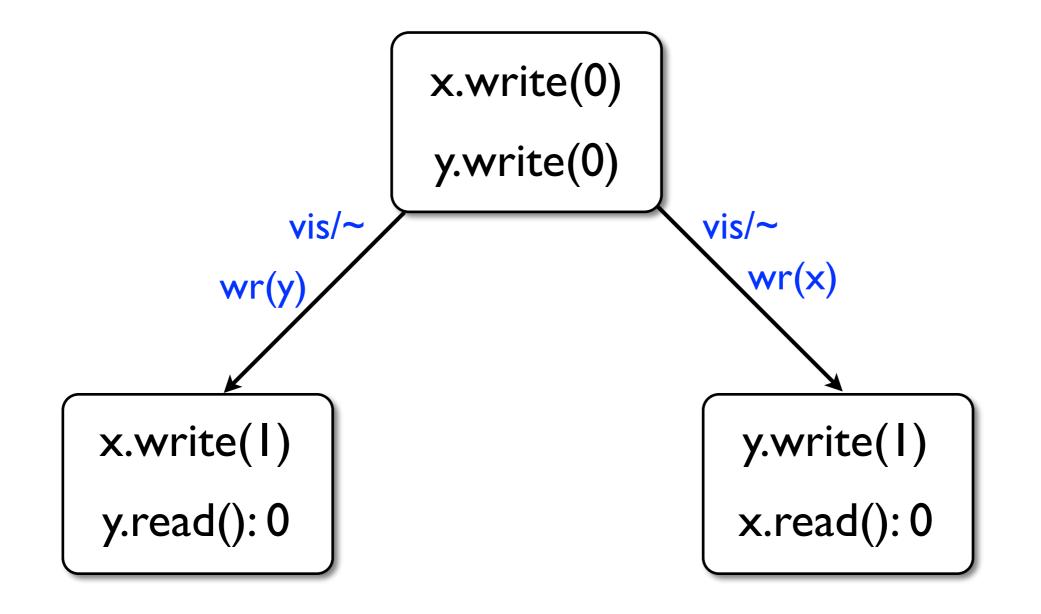
PSI allows only cycles in (wr \cup ww \cup wr) with at least two distinct rw edges

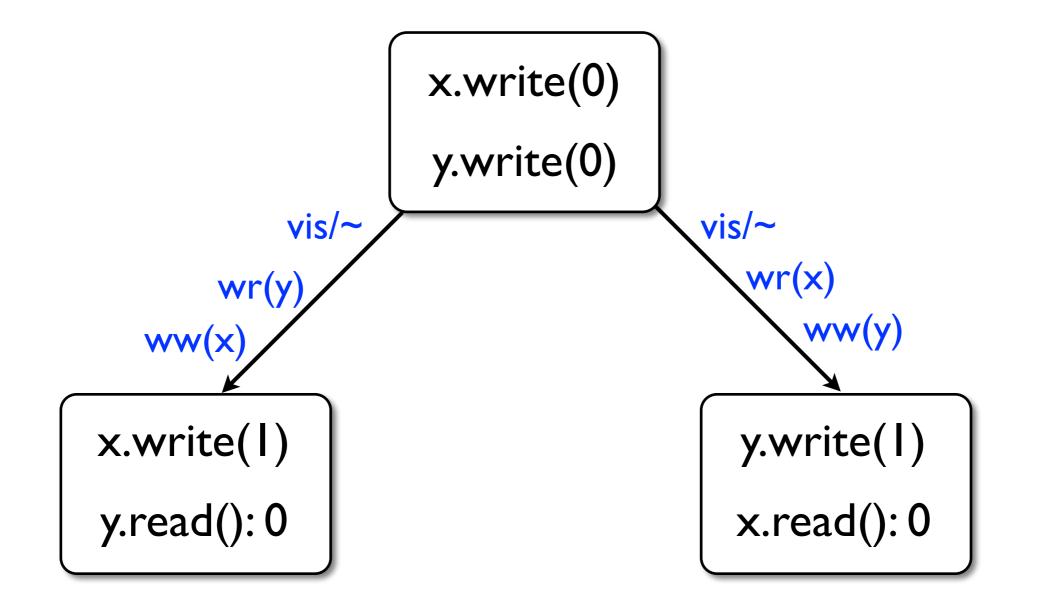
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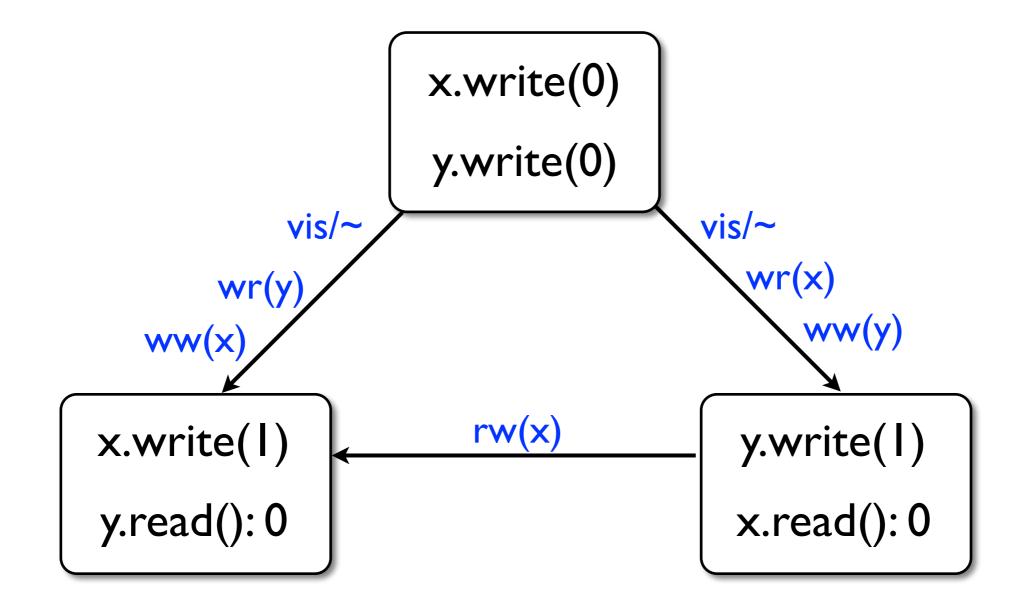
If (wr \cup ww \cup wr) for a PSI execution contains a cycle, then it also contains one:

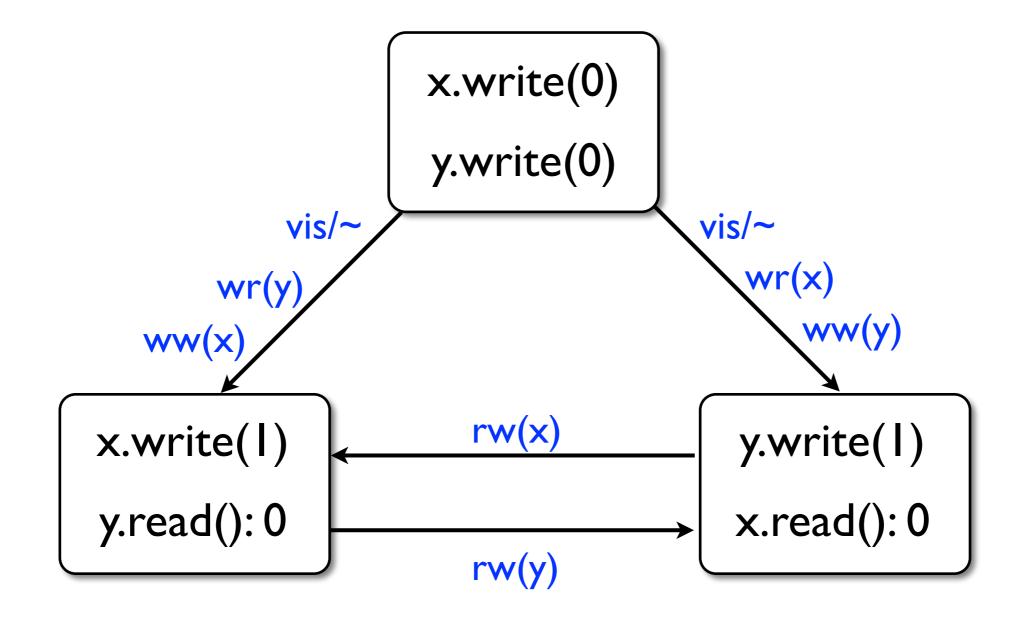
- with at least two rw edges, and
- where all rw edges are due to distinct objects

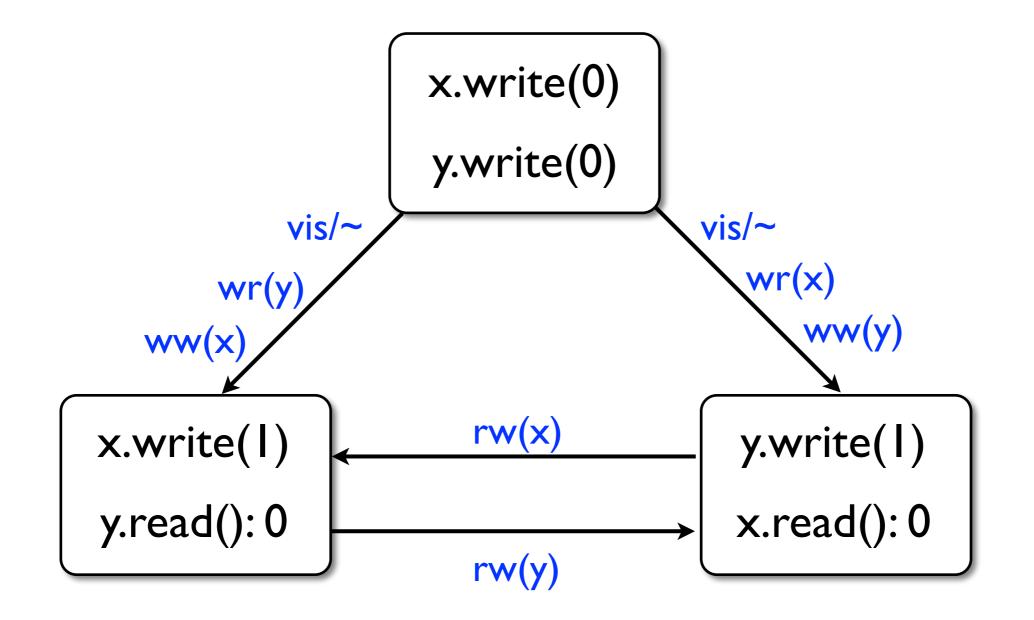




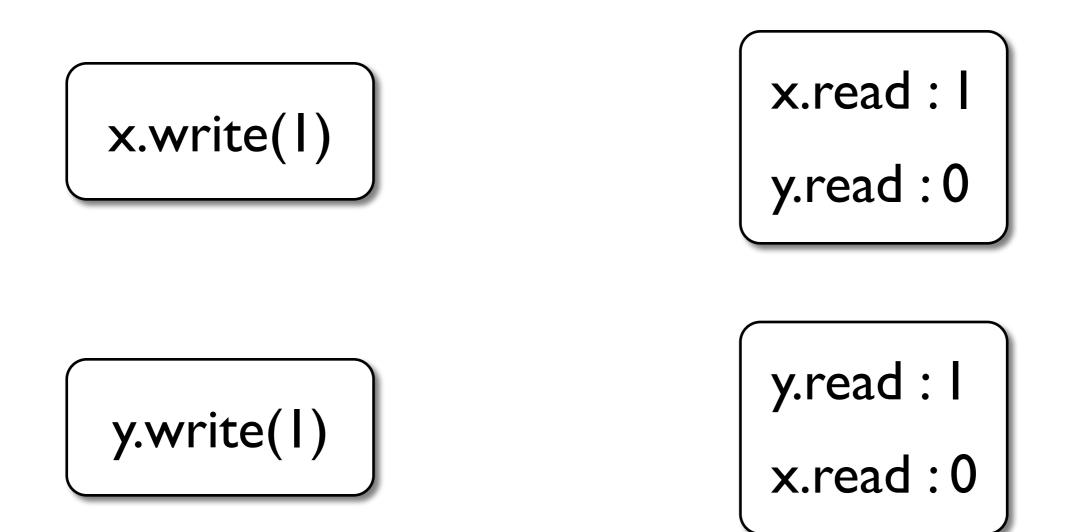


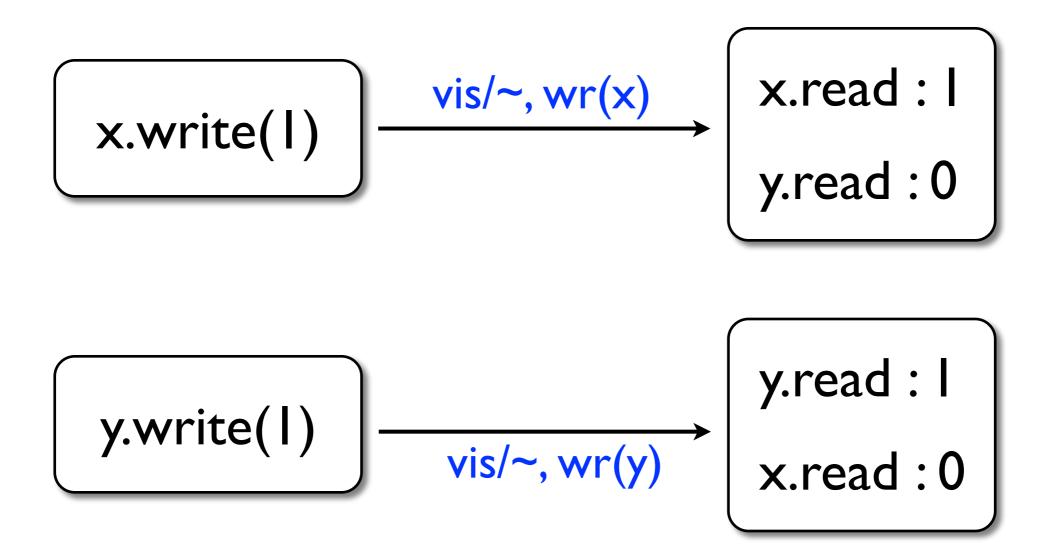


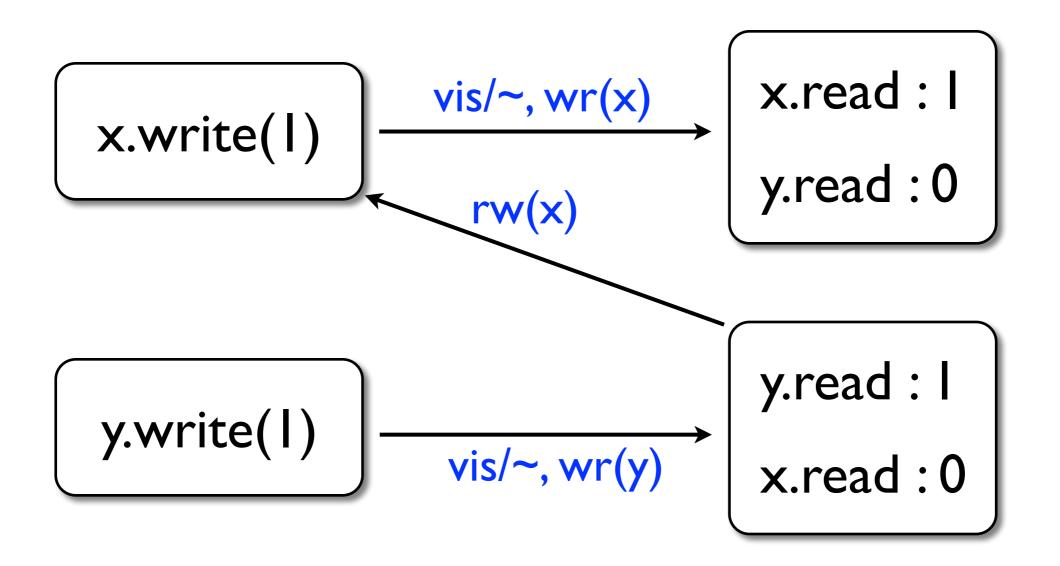


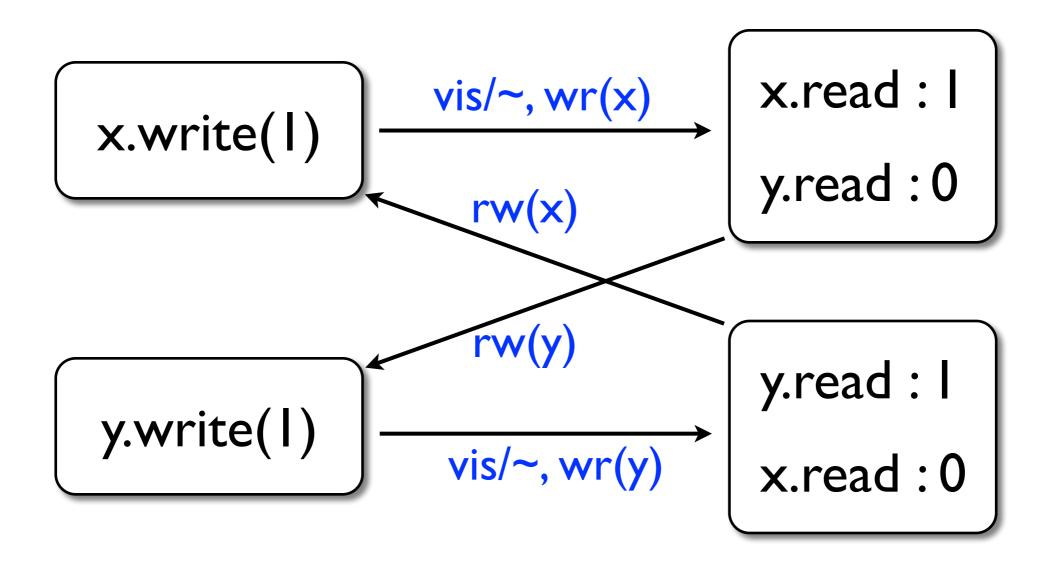


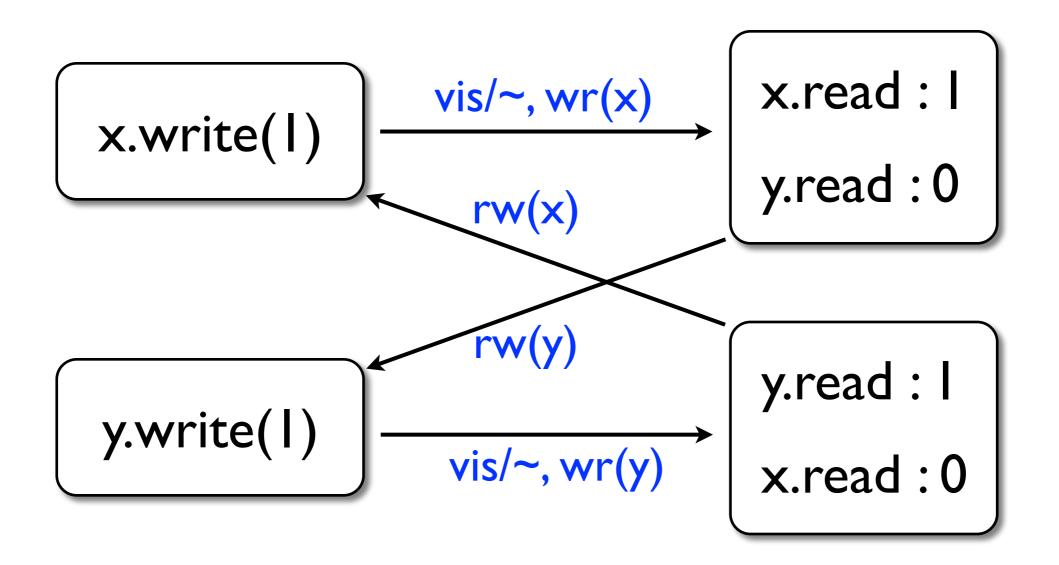
Cycle with 2 rw on different objects: allowed by PSI



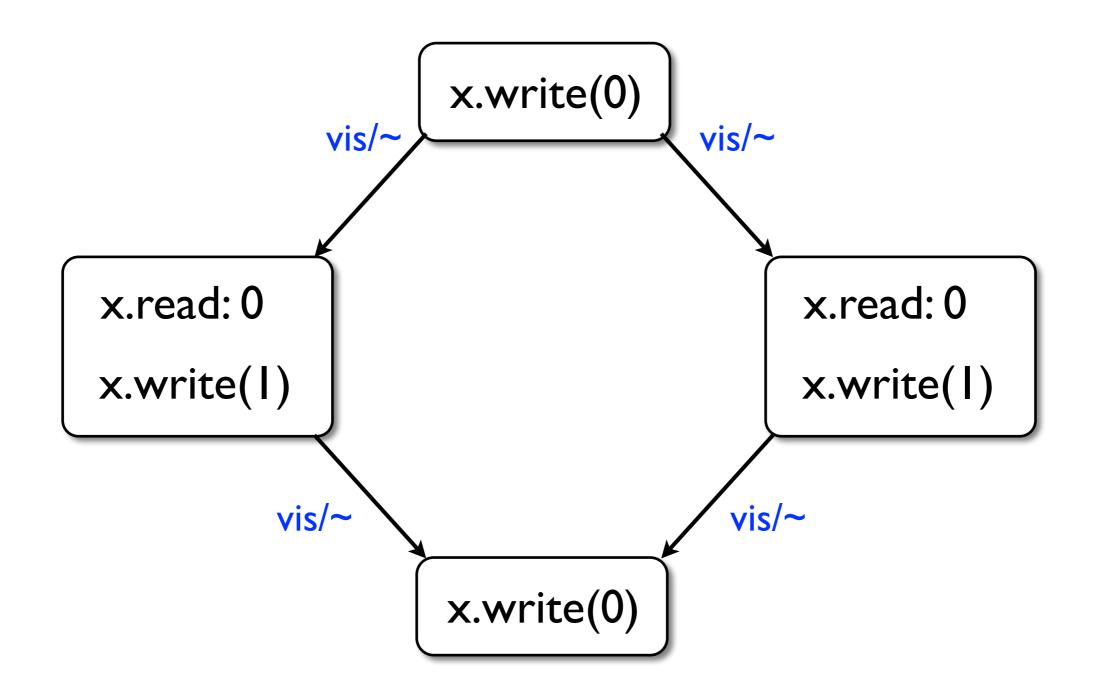


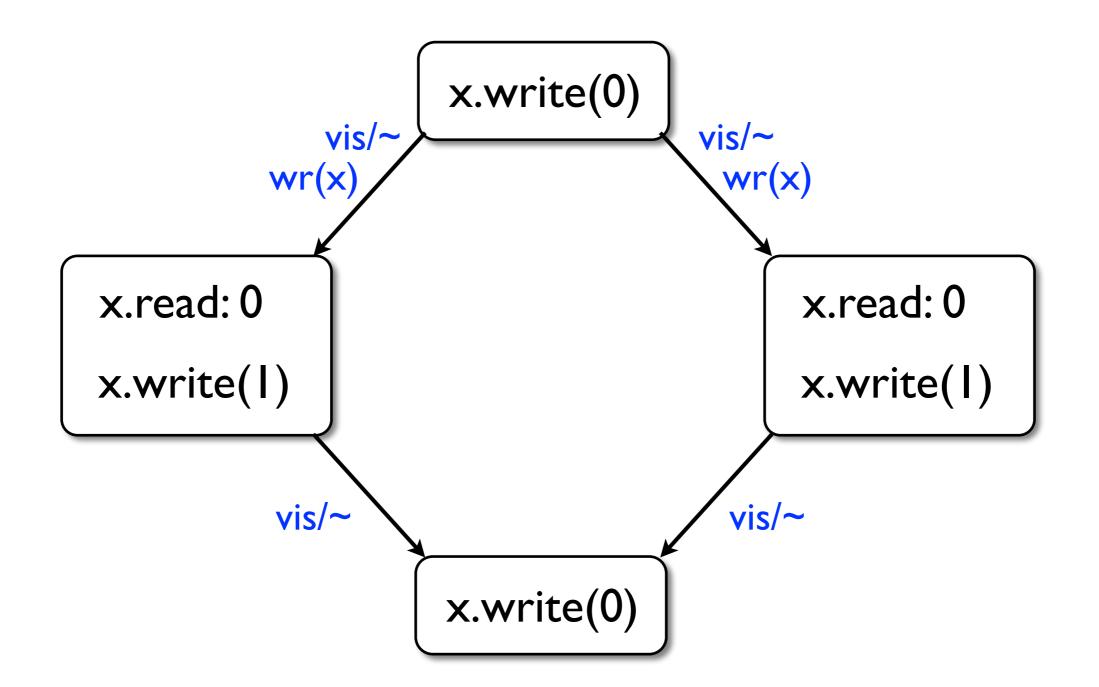


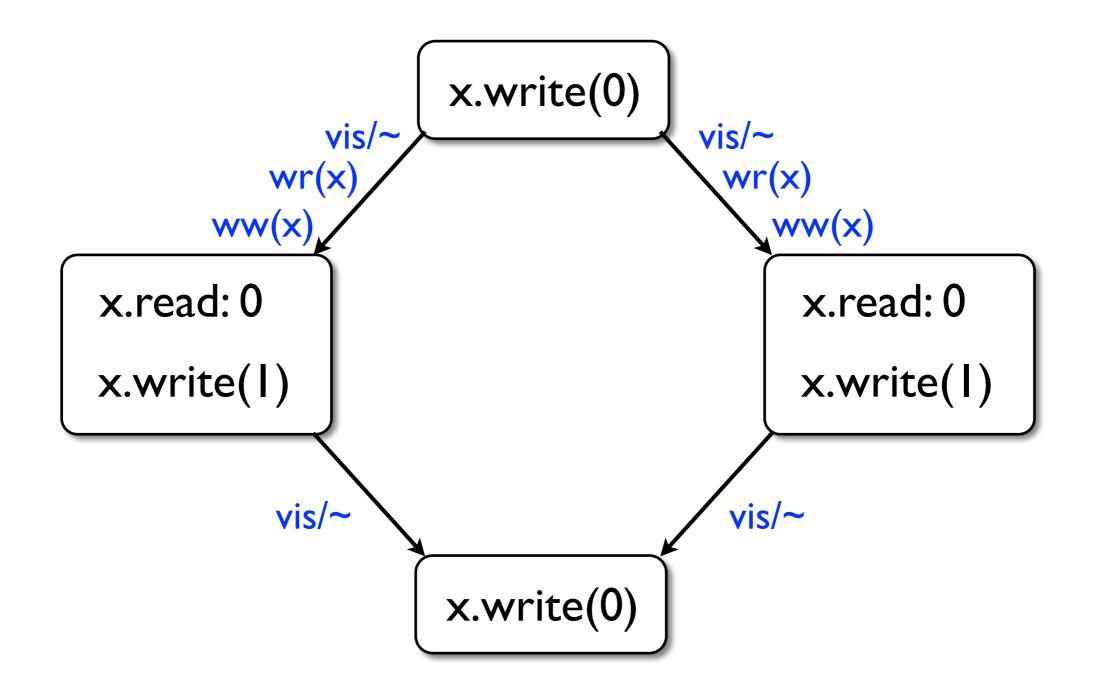


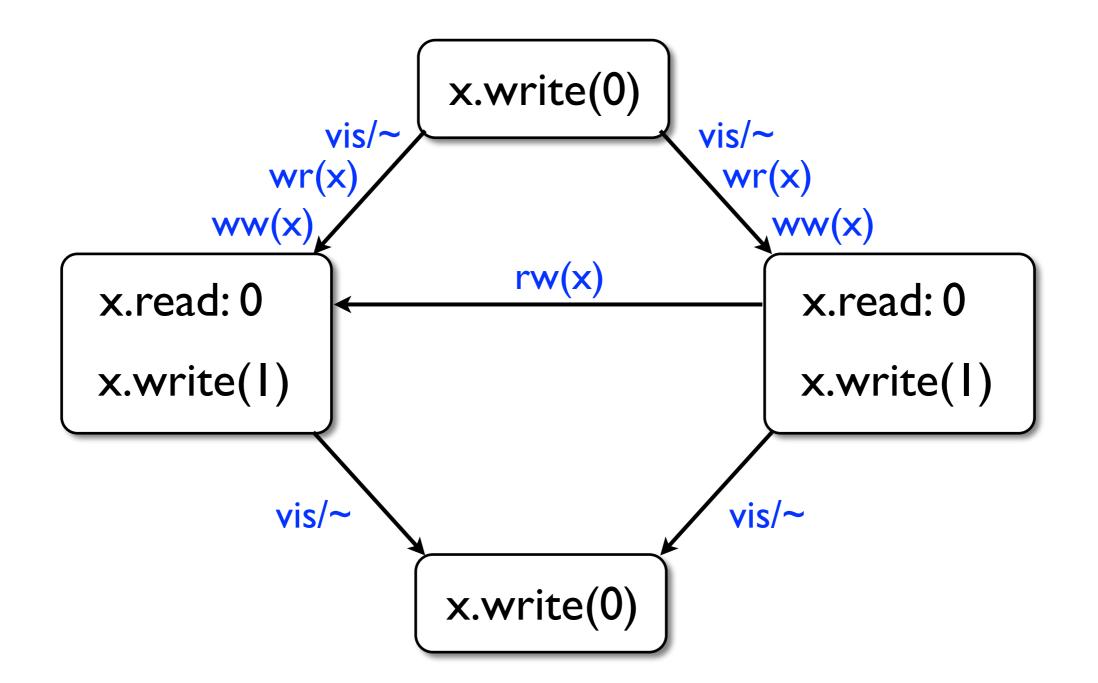


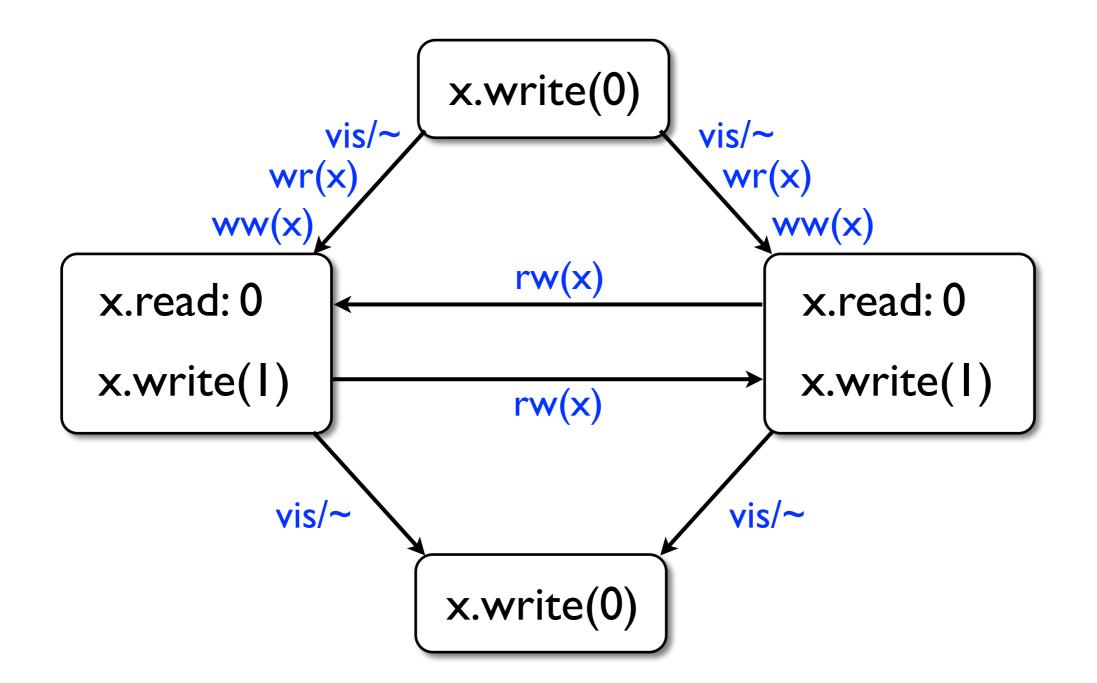
Cycle with 2 rw on different objects: allowed by PSI

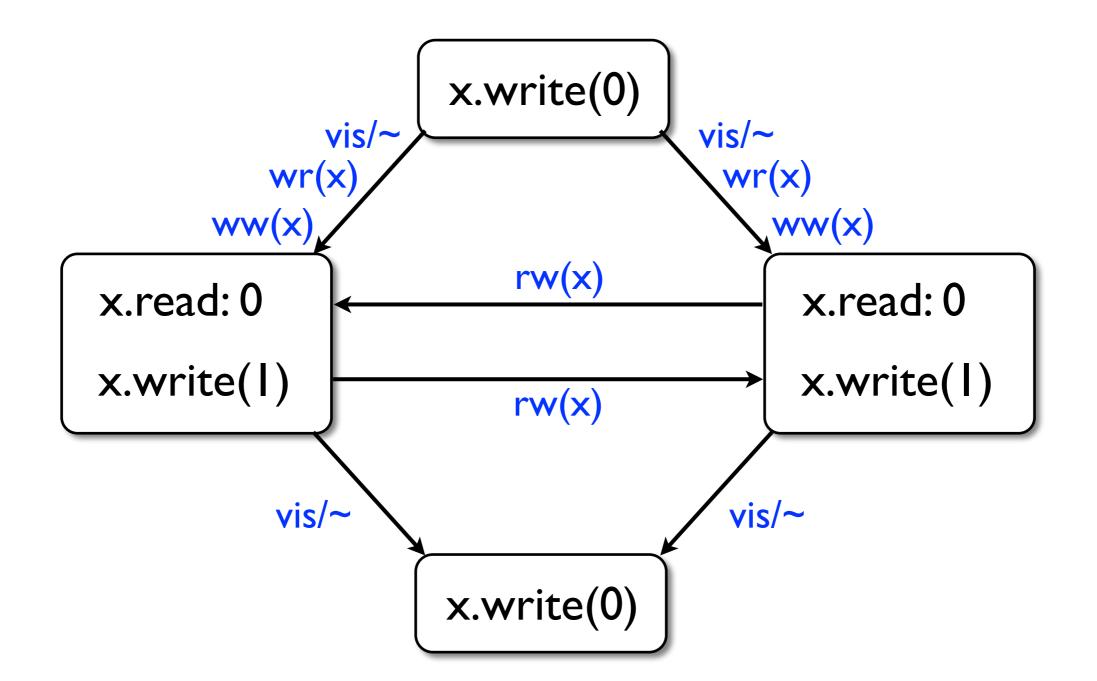








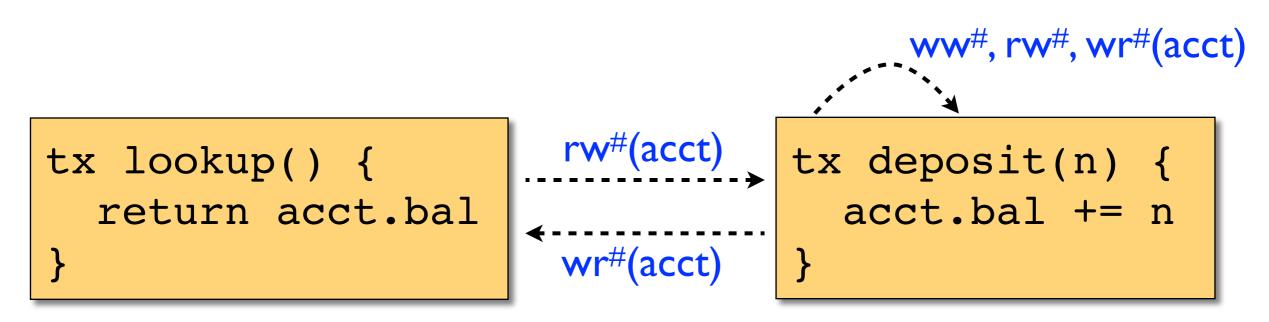




Not a valid PSI execution: violates write-conflict detection The 2 rw edges are due to the same object

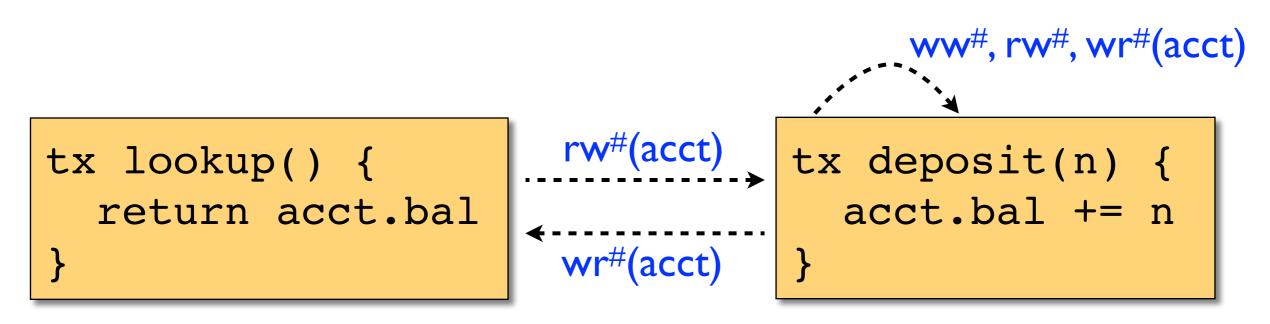
If a dependency graph of a PSI execution contains a cycle, then it also contains one:

- with at least two rw edges, and
- where all rw edges are due to distinct objects



If a dependency graph of a PSI execution contains a cycle, then it also contains one:

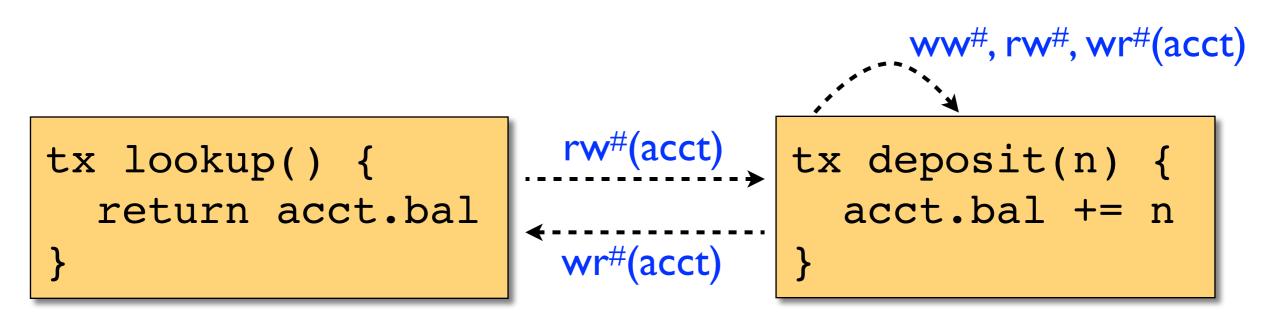
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No cycles in $wr^{\#} \cup ww^{\#} \cup rw^{\#}$ with all $rw^{\#}$ on different objects

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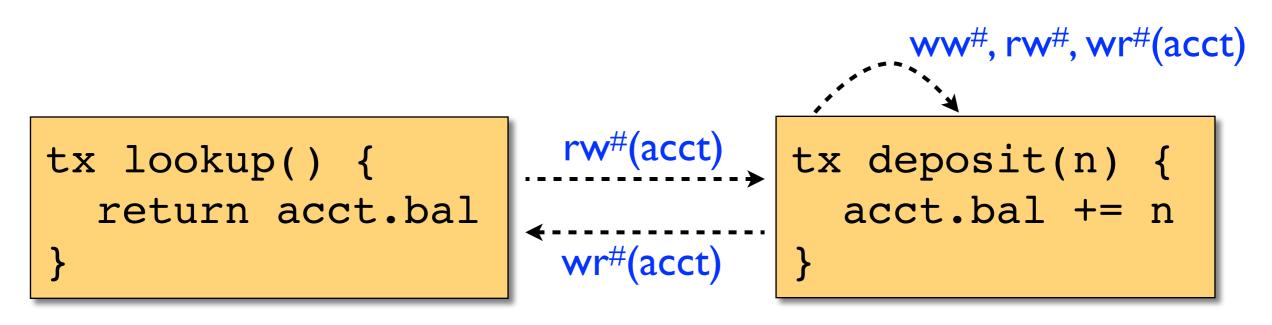
- with at least two rw edges, and
- where all rw edges are due to distinct objects



No cycles in wr[#] \cup ww[#] \cup rw[#] with all rw[#] on different objects \implies no such cycles in wr \cup ww \cup rw

If a dependency graph of a PSI execution contains a cycle, then it also contains one:

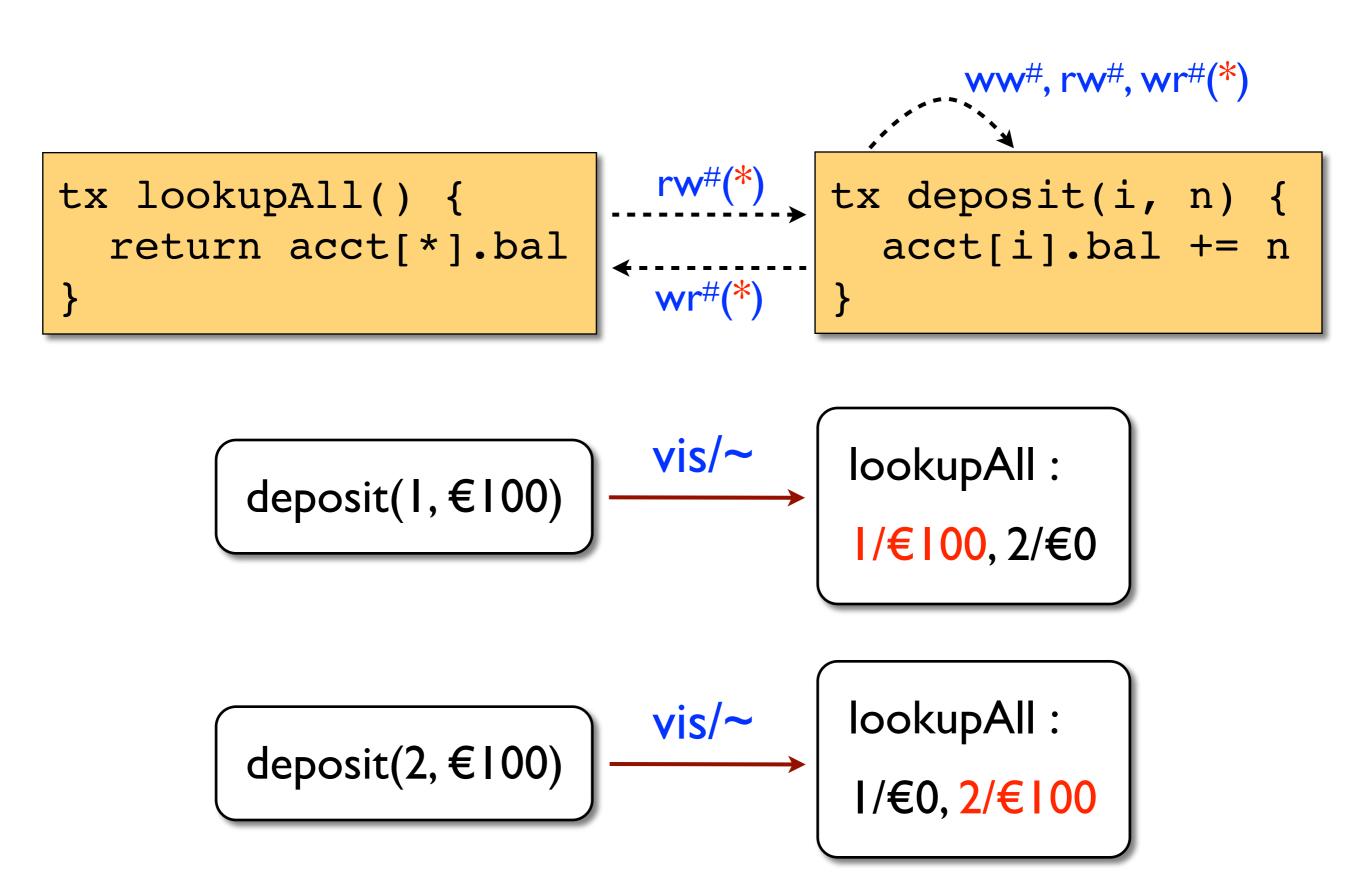
- with at least two rw edges, and
- where all rw edges are due to distinct objects



No cycles in $wr^{\#} \cup ww^{\#} \cup rw^{\#}$ with all $rw^{\#}$ on different objects

- \implies no such cycles in wr \cup ww \cup rw
- \implies application is serializable

Non-robustness



Automatic robustness checking

• Methods for other consistency models are similar

- Basis for practical tools [Warszawski et al., SIGMOD'17, Brutschy et al., PLDI'18; Nagar et al., CONCUR'18]
- Static criterion on graphs sometimes used to prune the search space before a more expensive analysis with more semantic information
- Can be used for bug-finding in the absence of specifications

Automatic robustness checking

ACIDRain: Concurrency-Related Attacks on Database-Backed Web Applications

Todd Warszawski, Peter Bailis Stanford InfoLab

ABSTRACT

In theory, database transactions protect application data from corruption and integrity violations. In practice, database transactions frequently execute under weak isolation that exposes programs to a range of concurrency anomalies, and programmers may fail to correctly employ transactions. While low transaction volumes mask many potential concurrency-related errors under normal operation, determined adversaries can exploit them programmatically for fun and profit. In this paper, we formalize a new kind of attack on database-backed applications called an ACIDRain attack, in which an adversary systematically exploits concurrency-related vulnerabilities via programmatically accessible APIs. These attacks are not theoretical: ACIDRain attacks have already occurred in a handful of applications in the wild, including one attack which bankrupted a popular Bitcoin exchange. To proactively detect the potential for ACIDRain attacks, we extend the theory of weak isolation to analyze latent potential for non-serializable behavior under concurrent web API calls. We introduce a language-agnostic method for detecting potential isolation anomalies in web applications, called Abstract Anomaly Detection (2AD), that uses dynamic traces of database accesses to efficiently reason about the space of possible concurrent interleavings. We apply a prototype 2AD analysis tool to 12 popular self-hosted eCommerce applications written in four languages and

1 2 3 4	<pre>def withdraw(amt, user_id): bal = readBalance(user_id) if (bal >= amt): writeBalance(bal - amt, user_id)</pre>	(a)
1	def withdraw(amt, user_id):	(b)
2	beginTxn()	
2 3	beginTxn() bal = readBalance(user_id)	
2 3 4	0	
I	$bal = readBalance(user_id)$	

Figure 1: (a) A simplified example of code that is vulnerable to an ACIDRain attack allowing overdraft under concurrent access. Two concurrent instances of the withdraw function could both read balance 100, check that $100 \ge 99$, and each allow 99 to be withdrawn, resulting 198 total withdrawals. (b) Example of how transactions could be inserted to address this error. However, even this code is vulnerable to attack at isolation levels at or below Read Committed, unless explicit locking such as SELECT_EOR_UPDATE is used. While this scenario closely re-

Implementing strong consistency

Designing consistency protocols

 So far implementations have been lightweight: "an operation can only be delivered after all its causal dependencies"

 In reality, designing consistency protocols and proving them correct is very difficult!

• Even more so for strong consistency protocols



c.withdraw(100): ?



c.withdraw(100) : ?

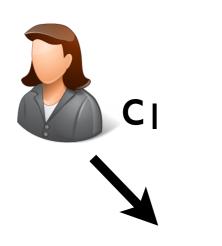


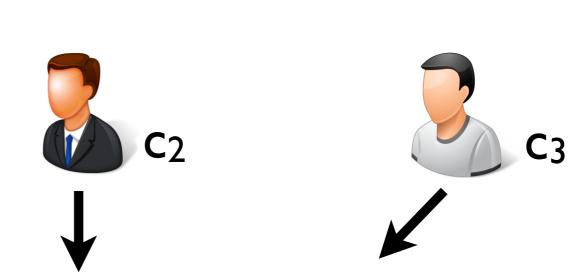




c.withdraw(100) : ?

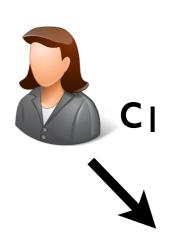
Sombody has to order commands

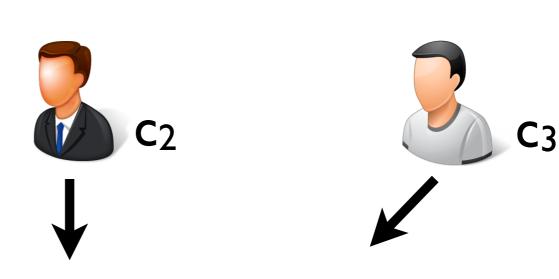






Single server, clients send commands to the server



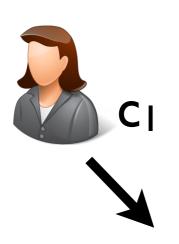


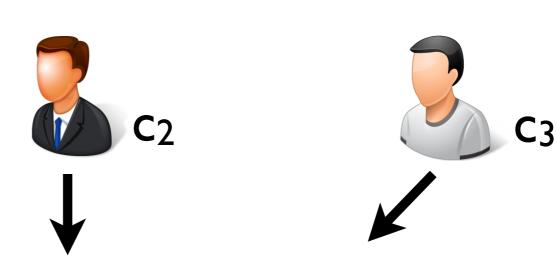
C1, C2, C3



r₁, r₂, r₃

Server totally orders commands and computes the sequence of results



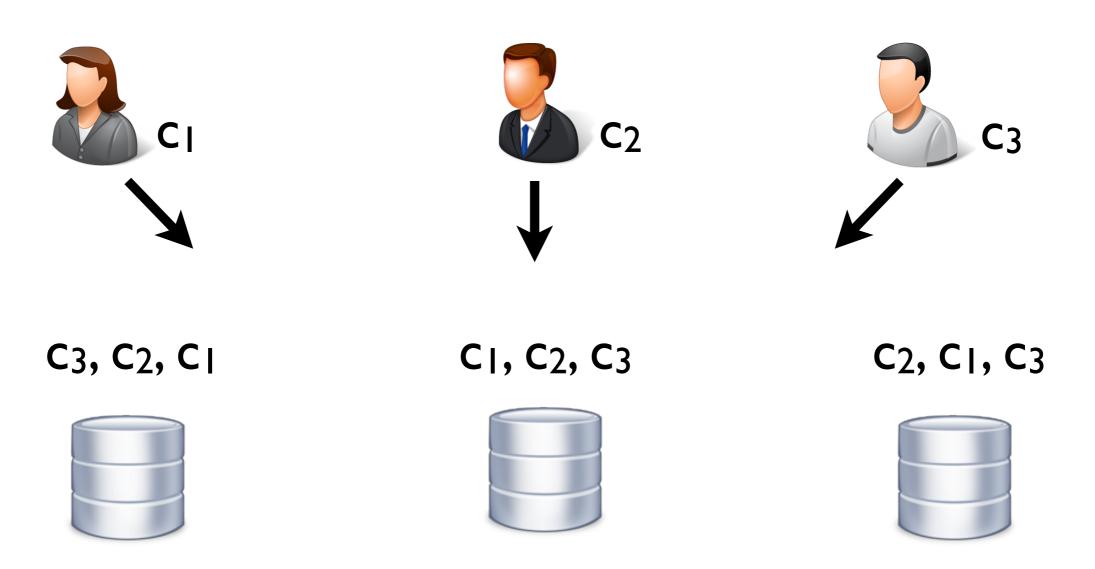


C1, C2, C3

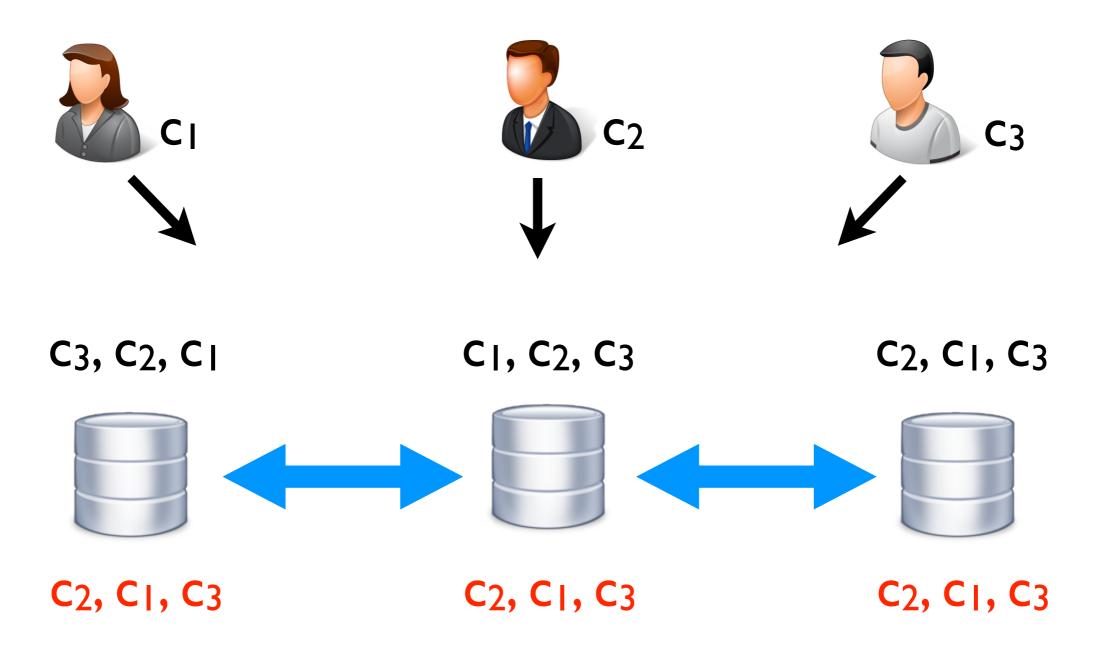


r₁, r₂, r₃

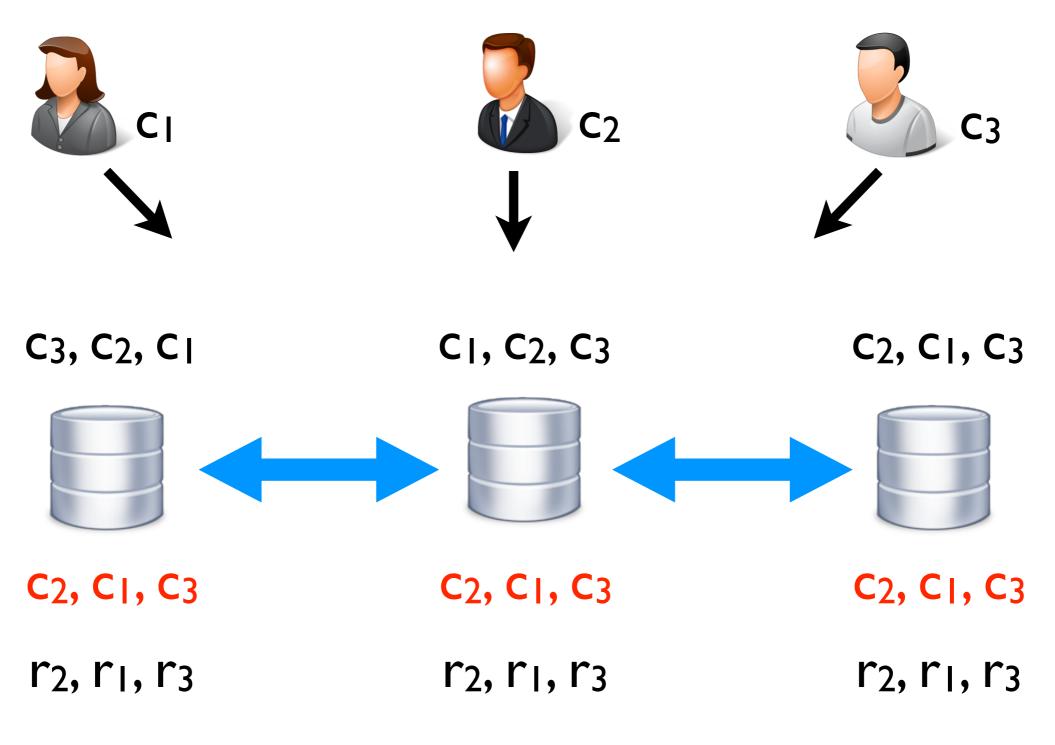
Servers can crash! Need a fault-tolerant solution



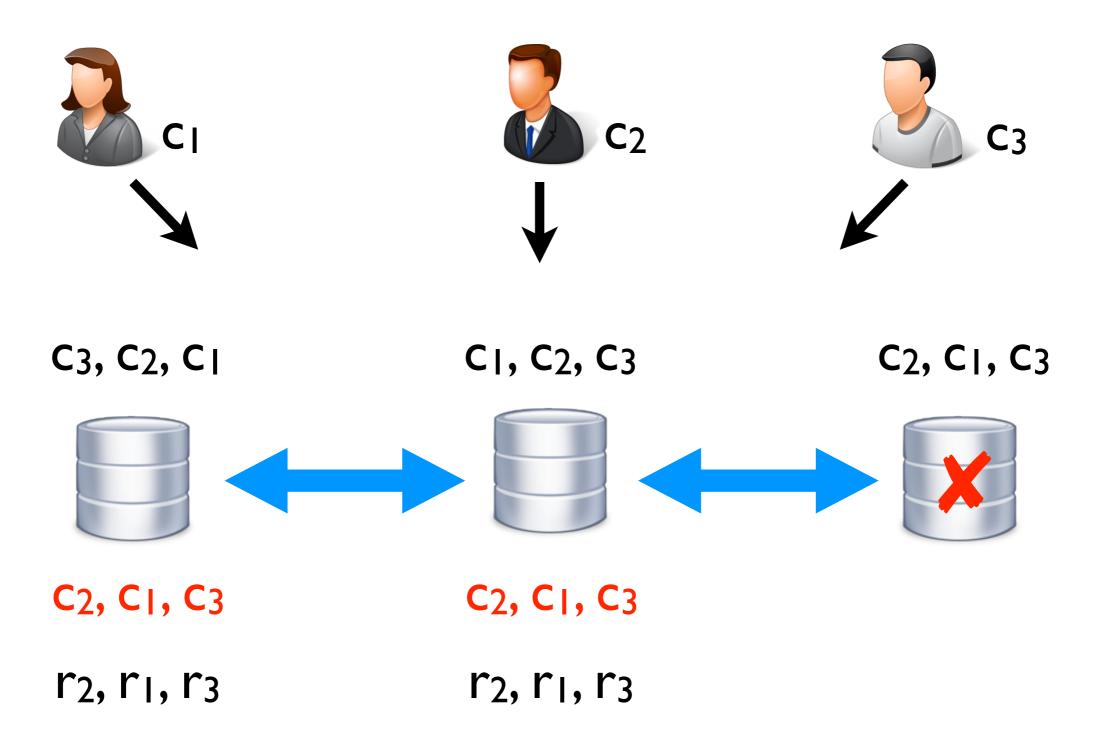
Clients send commands to all replicas Replicas may receive commands in different orders



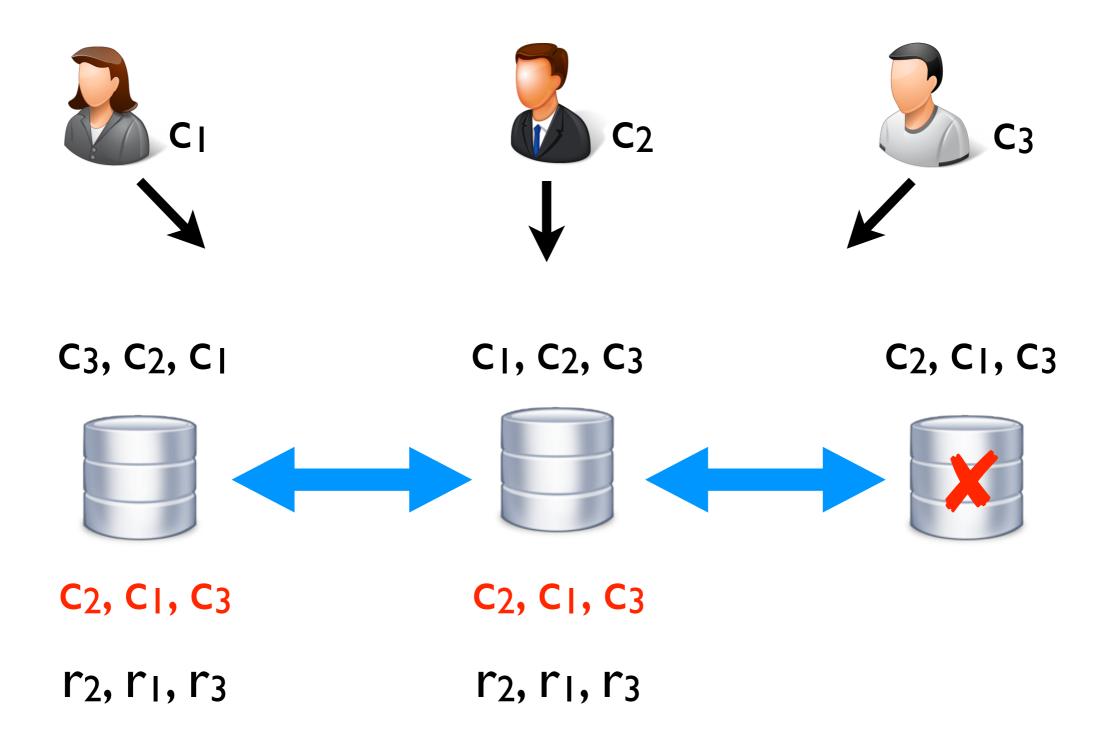
A distributed protocol totally order commands: needs synchronisation



Operations are deterministic \implies replicas compute the same sequence of results

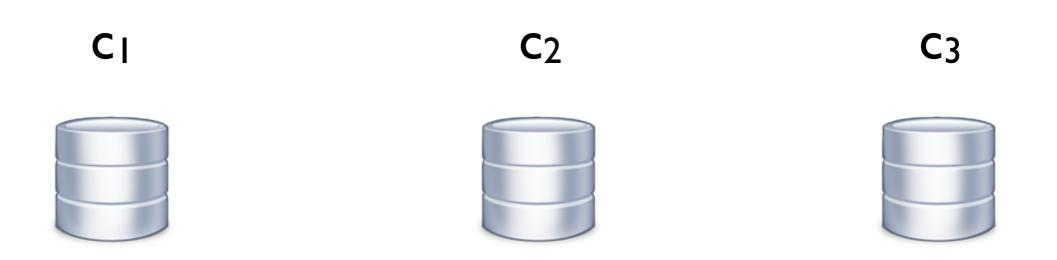


Implements sequential consistency (in fact, linearizability)



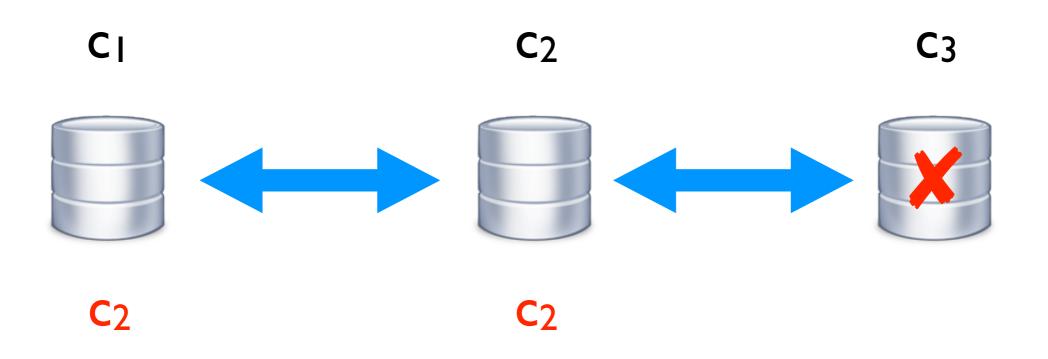
SMR requires solving a sequence of consensus instances: agree on the next command to execute

Consensus



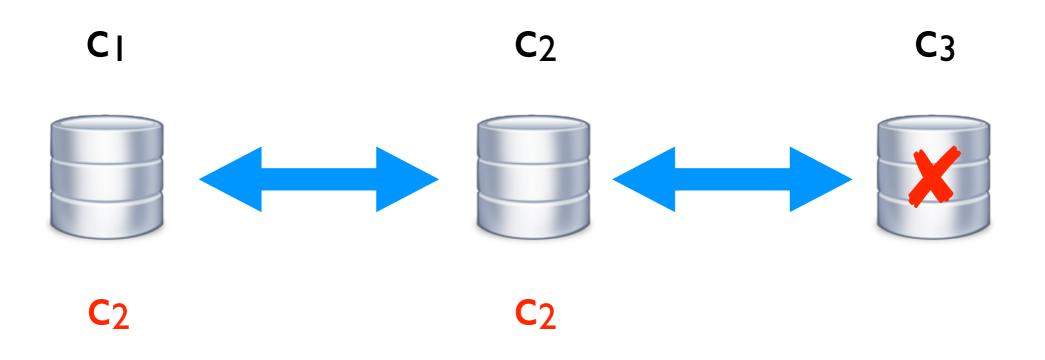
- Several nodes, which can crash
- Each proposes a value

Consensus



- Several nodes, which can crash
- Each proposes a value
- All non-crashed nodes agree on a single value

Consensus



- Challenge: asynchronous channels ⇒
 can't tell a crashed node from a slow one!
- Assume only a minority of nodes can crash: a majority reach an agreement

The zoo of consensus protocols

- Viewstamped replication (1988)
- Paxos (1998)
- Disk Paxos (2003)
- Cheap Paxos (2004)
- Generalized Paxos (2004)
- Paxos Commit (2004)
- Fast Paxos (2006)
- Stoppable Paxos (2008)

- Mencius (2008)
- Vertical Paxos (2009)
- ZAB (2009)
- Ring Paxos (2010)
- Egalitarian Paxos (2013)
- Raft (2014)
- M2Paxos (2016)
- Flexible Paxos (2016)
- Caesar (2017)

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The zoo of co

Complex protocols: constant fight for better performance

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LESLIE LAMPORT Digital Equipment Corporation

Recent archaeological discoveries on the island of Paxos reveal that the parliament functioned despite the peripatetic propensity of its part-time legislators. The legislators maintained consistent copies of the parliamentary record, despite their frequent forays from the chamber and the forgetfulness of their messengers. The Paxon parliament's protocol provides a new way of implementing the state-machine approach to the design of distributed systems.

Categories and Subject Descriptors: C2.4 [Computer-Communications Networks]: Distributed Systems—*Network operating systems*; D4.5 [Operating Systems]: Reliability—*Fault-tolerance*; J.1 [Administrative Data Processing]: Government

General Terms: Design, Reliability

Additional Key Words and Phrases: State machines, three-phase commit, voting

This submission was recently discovered behind a filing cabinet in the *TOCS* editorial office. Despite its age, the editor-in-chief felt that it was worth publishing. Because the author is currently doing field work in the Greek isles and cannot be reached, I was asked to prepare it for publication.

The author appears to be an archeologist with only a passing interest in computer science. This is unfortunate; even though the obscure ancient Paxon civilization he describes is of little interest to most computer scientists, its legislative system is an excellent model for how to implement a distributed computer system in an asynchronous environment. Indeed, some of the refinements the Paxons made to their protocol appear to be unknown in the systems literature.

LESLIE LAMPORT Digital Equipment Corporation

Paxos Made Simple

Leslie Lamport

Abstract

The Paxos algorithm, when presented in plain English, is very simple.

LESLIE LAMPORT Digital Equipment Corporation

Paxos Made Simple

Paxos Made Moderately Complex

ROBBERT VAN RENESSE and DENIZ ALTINBUKEN, Cornell University

This article explains the full reconfigurable multidecree Paxos (or multi-Paxos) protocol. Paxos is by no means a simple protocol, even though it is based on relatively simple invariants. We provide pseudocode and explain it guided by invariants. We initially avoid optimizations that complicate comprehension. Next we discuss liveness, list various optimizations that make the protocol practical, and present variants of the protocol.

Categories and Subject Descriptors: C.2.4 [Computer-Communication Networks]: Distributed Systems—Network operating systems; D.4.5 [Operating Systems]: Reliability—Fault-tolerance

General Terms: Design, Reliability

Additional Key Words and Phrases: Replicated state machines, consensus, voting

ACM Reference Format:

Robbert van Renesse and Deniz Altinbuken. 2015. Paxos made moderately complex. ACM Comput. Surv. 47, 3, Article 42 (February 2015), 36 pages. DOI: http://dx.doi.org/10.1145/2673577

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Paxos Made Simple

Paxos Made Moderately Complex

In Search of an Understandable Consensus Algorithm

Diego Ongaro and John Ousterhout Stanford University

Abstract

Raft is a consensus algorithm for managing a replicated log. It produces a result equivalent to (multi-)Paxos, and it is as efficient as Paxos, but its structure is different from Paxos; this makes Raft more understandable than Paxos and also provides a better foundation for building practical systems. In order to enhance understandability, Raft separates the key elements of consensus, such as leader election, log replication, and safety, and it enforces a stronger degree of coherency to reduce the number of states that must be considered. Results from a user study demonstrate that Raft is easier for students to learn than Paxos. Raft also includes a new mechanism for changing the cluster membership, which uses overlapping majorito understand than Paxos: after learning both algorithms, 33 of these students were able to answer questions about Raft better than questions about Paxos.

Raft is similar in many ways to existing consensus algorithms (most notably, Oki and Liskov's Viewstamped Replication [27, 20]), but it has several novel features:

- Strong leader: Raft uses a stronger form of leadership than other consensus algorithms. For example, log entries only flow from the leader to other servers. This simplifies the management of the replicated log and makes Raft easier to understand.
- Leader election: Raft uses randomized timers to elect leaders. This adds only a small amount of mechanism to the heartbeats already required for any consensus al-

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Unfortunately, Paxos has two significant drawbacks. The first drawback is that Paxos is exceptionally difficult to understand. The full explanation [15] is notoriously opaque; few people succeed in understanding it, and only with great effort. As a result, there have been several attempts to explain Paxos in simpler terms [16, 20, 21]. These explanations focus on the single-decree subset, yet they are still challenging. In an informal survey of attendees at NSDI 2012, we found few people who were comfortable with Paxos, even among seasoned researchers. We struggled with Paxos ourselves; we were not able to understand the complete protocol until after reading several simplified explanations and designing our own alter-

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Paxos Made Live - An Engineering Perspective (2006 Invited Talk)

Tushar Chandra, Robert Griesemer, and Joshua Redstone

Google Inc.

ABSTRACT

We describe our experience in building a fault-tolerant database using the Paxos consensus algorithm. Despite the existing literature in the field, building such a database proved to be non-trivial. We describe selected algorithmic and engineering problems encountered, and the solutions we found for them. Our measurements indicate that we have built a competitive system.

Categories and Subject Descriptors

D.4.5 [Operating systems]: Reliability—Fault-tolerance; B.4.5 [Input/output and data communications]: Reliability, Testing, and Fault-Tolerance—Redundant design

General Terms

Experimentation, Performance, Reliability

Keywords

Experiences, Fault-tolerance, Implementation, Paxos

database is just an example. As a result, the consensus problem has been studied extensively over the past two decades. There are several well-known consensus algorithms that operate within a multitude of settings and which tolerate a variety of failures. The Paxos consensus algorithm [8] has been discussed in the theoretical [16] and applied community [10, 11, 12] for over a decade.

We used the Paxos algorithm ("Paxos") as the base for a framework that implements a fault-tolerant log. We then relied on that framework to build a fault-tolerant database. Despite the existing literature on the subject, building a production system turned out to be a non-trivial task for a variety of reasons:

• While Paxos can be described with a page of pseudocode, our complete implementation contains several thousand lines of C++ code. The blow-up is not due simply to the fact that we used C++ instead of pseudo notation, nor because our code style may have been verbose. Converting the algorithm into a practical, production-ready system involved implementing many features and optimizations – some published in the lit-

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• There are significant gaps between the description of the Paxos algorithm and the needs of a real-world system. In order to build a real-world system, an expert needs to use numerous ideas scattered in the literature and make several relatively small protocol extensions. The cumulative effort will be substantial and the final system will be based on an unproven protocol.

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Paxos Made Live - 5.1 Handling disk corruption (20(

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Replicas witness disk corruption from time to time. A disk may be corrupted due to a media failure or due to an operator error (an operator may accidentally erase critical data). When a replica's disk is corrupted and it loses its Tushar Chandra, Rol persistent state, it may renege on promises it has made to other replicas in the past. This violates a key assumption in the Paxos algorithm. We use the following mechanism to address this problem [14].

> Disk corruptions manifest themselves in two ways. Either file(s) contents may change or file(s) may become inaccessible. To detect the former, we store the checksum of the contents of each file in the file². The latter may be indistinguishable from a new replica with an empty disk – we detect this case by having a new replica leave a marker in GFS after start-up. If this replica ever starts again with an empty disk, it will discover the GFS marker and indicate that it has a corrupted disk.

> A replica with a corrupted disk rebuilds its state as follows. It participates in Paxos as a non-voting member; meaning that it uses the catch-up mechanism to catch up but does not respond with promise or acknowledgment messages. It remains in this state until it observes one complete instance of Paxos that was started after the replica started rebuilding its state. By waiting for the extra instance of Paxos, we ensure that this replica could not have reneged on an earlier promise.

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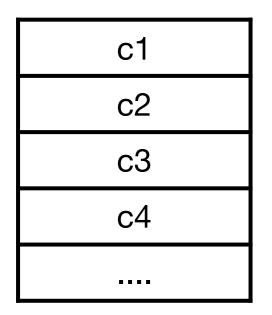
> Disk corruptions manifest themselves in two ways. Either file(s) contents may change or file(s) may become inaccessible. To detect the former, we store the checksum of the contents of each file in the file². The latter may be indistinguishable from a new replica with an empty disk – we detect this case by having a new replica leave a marker in

GFS a empty Broken [Michael et al., DISC'16] that it

A replica with a corrupted disk rebuilds its state as follows. It participates in Paxos as a non-voting member; meaning that it uses the catch-up mechanism to catch up but does not respond with promise or acknowledgment messages. It remains in this state until it observes one complete instance of Paxos that was started after the replica started rebuilding its state. By waiting for the extra instance of Paxos, we ensure that this replica could not have reneged on an earlier promise.

> production-ready system involved implementing many features and optimizations – some published in the lit-

Another application: blockchain





- Blockchain = using consensus to agree on a sequence of blocks in a ledger
- Tolerates malicious behaviour: some nodes may deviate from the protocol
- Many protocols descended from Paxos



Facebook's Libra pitches to be the future of money



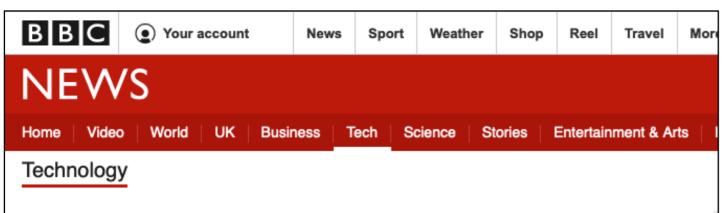
Rory Cellan-Jones Technology correspondent @BBCRoryCJ

() 18 June 2019



It is a hugely ambitious - some might say megalomaniacal - project to create a new global currency. Facebook's David Marcus tells me it is about giving billions of people more freedom with money and "righting the many wrongs of the present system".

The message is this is not some little side project a small team at the Facebook's



Facebook's Libra pitches to be the future

of mone [PODC'19]



HotStuff: BFT Consensus with Linearity and Responsiveness

() 18 June 2019



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The message is t

Maofan Yin Dahlia Malkhi Cornell University VMware Research

Michael K. Reiter UNC-Chapel Hill VMware Research Guy Golan Gueta VMware Research Ittai Abraham VMware Research

ABSTRACT

We present HotStuff, a leader-based Byzantine fault-tolerant replication protocol for the partially synchronous model. Once network communication becomes synchronous, HotStuff enables a correct leader to drive the protocol to consensus at the pace of actual (vs. maximum) network delay—a property called *responsiveness*—and with communication complexity that is linear in the number of replicas. To our knowledge, HotStuff is the first partially synchronous BFT replication protocol exhibiting these combined properties. Its simplicity enables it to be further pipelined and simplified into a practical, concise protocol for building large-scale replication services.

CCS CONCEPTS

Software and its engineering → Software fault tolerance;
 Security and privacy → Distributed systems security.

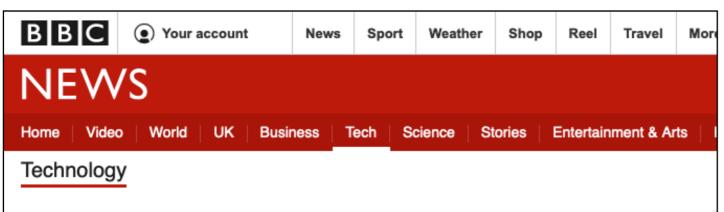
KEYWORDS

Byzantine fault tolerance; consensus; responsiveness; scalability; blockchain

stabilization time (GST). In this model, $n \ge 3f + 1$ is required for non-faulty replicas to agree on the same commands in the same order (e.g., [12]) and progress can be ensured deterministically only after GST [27].

When BFT SMR protocols were originally conceived, a typical target system size was n = 4 or n = 7, deployed on a local-area network. However, the renewed interest in Byzantine fault-tolerance brought about by its application to blockchains now demands solutions that can scale to much larger n. In contrast to *permissionless* blockchains such as the one that supports Bitcoin, for example, so-called *permissioned* blockchains involve a fixed set of replicas that collectively maintain an ordered ledger of commands or, in other words, that support SMR. Despite their permissioned nature, numbers of replicas in the hundreds or even thousands are envisioned (e.g., [30, 42]). Additionally, their deployment to wide-area networks requires setting Δ to accommodate higher variability in communication delays.

The scaling challenge. Since the introduction of PBFT [20], the first practical BFT replication solution in the partial synchrony model, numerous BFT solutions were built around its core two-



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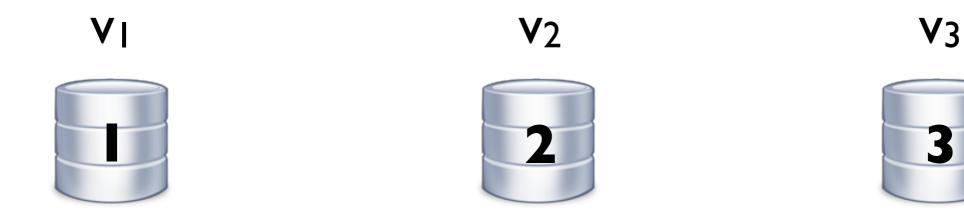
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ACKNOWLEDGMENTS

We are thankful to Mathieu Baudet, Avery Ching, George Danezis, François Garillot, Zekun Li, Ben Maurer, Kartik Nayak, Dmitri Perelman, and Ling Ren, for many deep discussions of HotStuff, and to Mathieu Baudet for exposing a subtle error in a previous version posted to the ArXiv of this manuscript.

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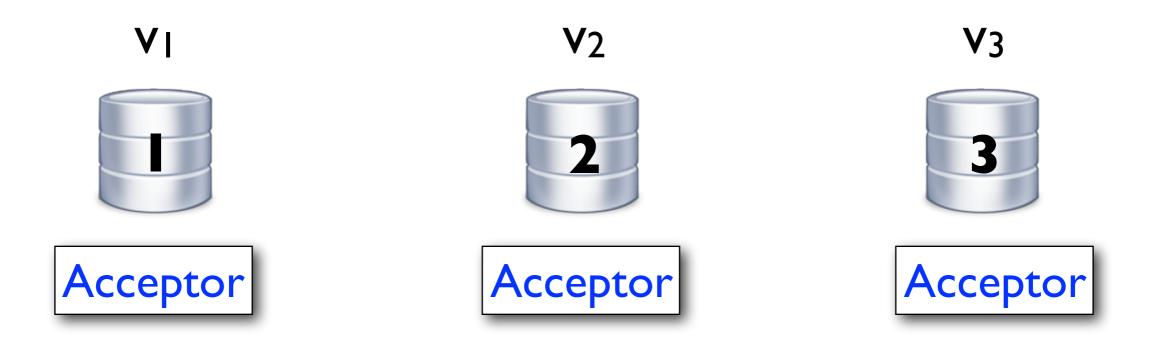
- 2f+1 nodes, at most f can crash
- Each node proposes a value
- All non-crashed nodes agree on a single value

VI

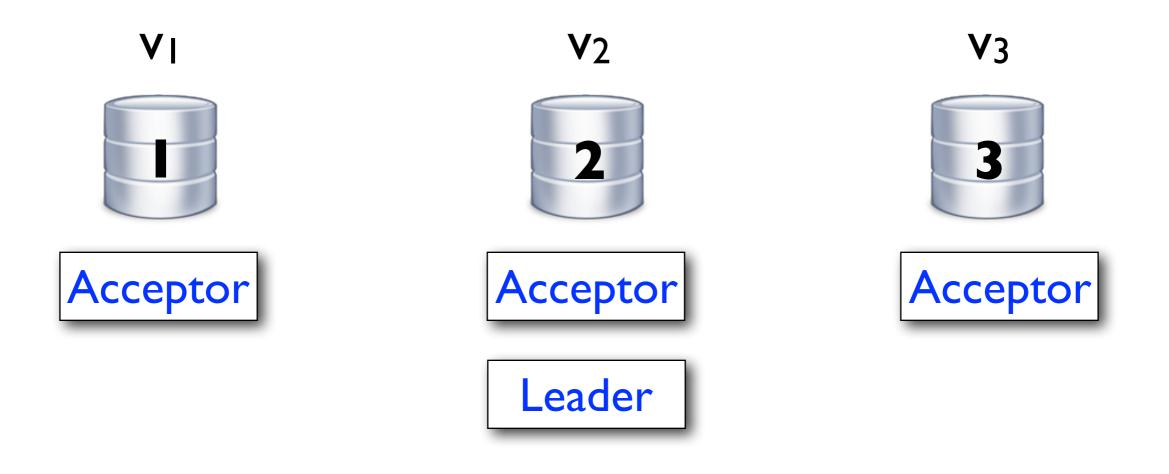








 Acceptors = members of parliament: can vote to accept a value, majority (quorum) wins



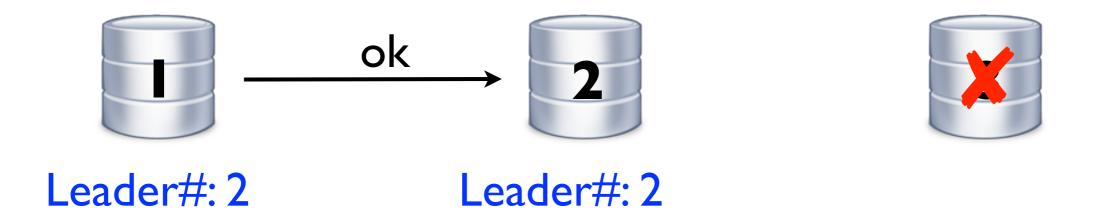
- Acceptors = members of parliament: can vote to accept a value, majority (quorum) wins
- Leader = parliament speaker: proposes its value to vote on
- Good for state-machine replication: can elect the leader once and get it to process multiple commands



Phase I: a prospective leader convinces a quorum of acceptors to accept its authority



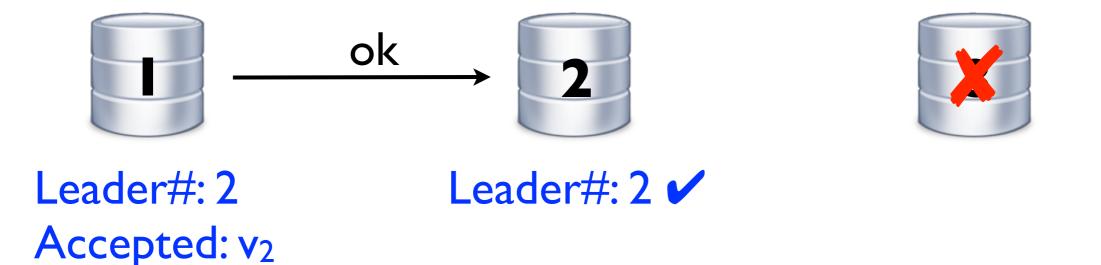
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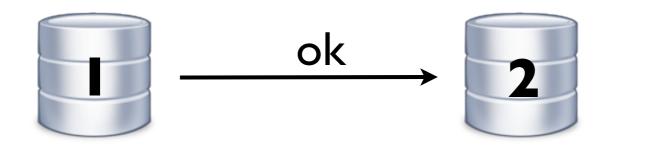














Leader#: 2 Accepted: v₂ Leader#: 2 Accepted: v₂

• Phase I: a prospective leader convinces a quorum of acceptors to accept its authority







Leader#: 2 Accepted: v₂

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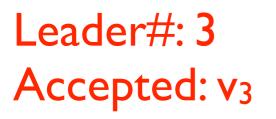


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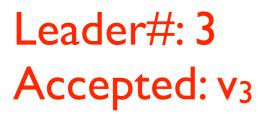


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Leader#: 3 Accepted: v₃ Reply v₃ to client

 Problem: node 3 may wake up, form a quorum of I and 3, and accept value v₃









Leader#: 2 ✓ Leader#: 3 ✓ Accepted: $v_2 \checkmark$ Accepted: $v_3 \checkmark$ Reply v₂ to client

Reply v_3 to client

- Problem: node 3 may wake up, form a quorum of I and 3, and accept value v_3
- Need to ensure once a value is chosen by a quorum, it can't be changed
- Use ballot numbers to distinguish different votes: unique for each potential leader



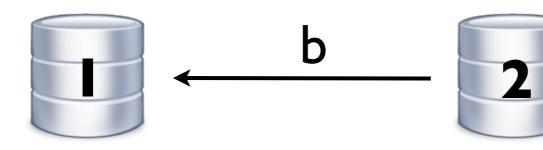
Leader#: ? Ballot#: 0 Accepted: ?



Leader#: ? Ballot#: 0 Accepted: ?



- Phase I: a prospective leader choses a ballot b and convinces a quorum of acceptors to switch to b
- Acceptor switches only if it's current ballot is smaller



Leader#: ? Ballot#: 0 Accepted: ? Leader#: 2 Ballot#: b Accepted: ?



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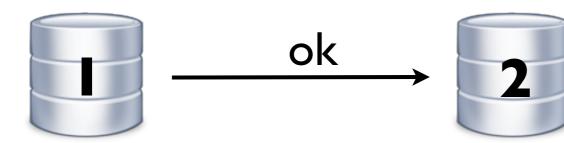
b, **v**₂



Leader#: 2 Ballot#: b Accepted: ?

Leader#: 2 Ballot#: b Accepted: v₂@b

- Phase 2: the leader sends its value tagged with its ballot number
- Acceptor only accepts a value tagged with the ballot it is in





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ok



Leader#: 2 Ballot#: b Accepted: $v_2@b$

Leader#: 2 V Ballot#: b Accepted: v₂@b Accepted: ? Reply v_2 to client

Leader#:? Ballot#:0

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Leader#: 2 Ballot#: b Accepted: v₂@b Reply v₂ to client





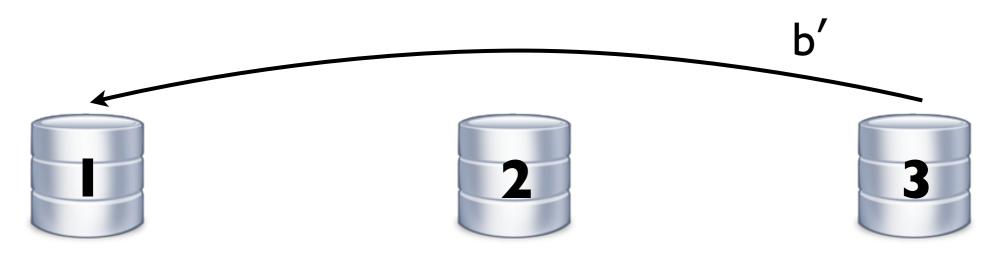


Leader#: 2 Ballot#: b Accepted: $v_2@b$

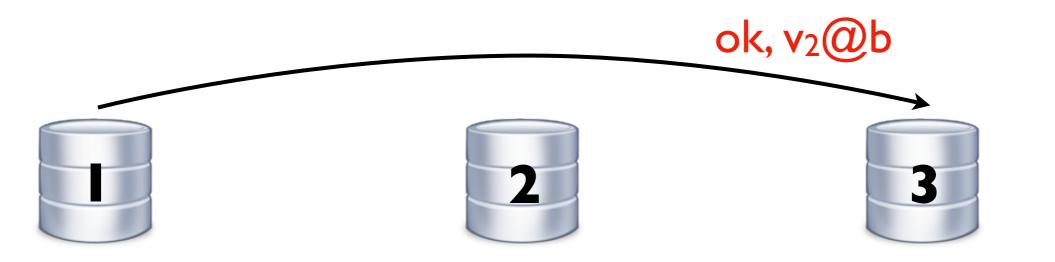
Leader#: 2 V Ballot#: b Accepted: v₂@b Accepted: ? Reply v_2 to client

Leader#:? Ballot#:0

- Need to ensure once a value is chosen by a quorum, it can't be changed
- Need do change Phase I to restrict which values can be proposed

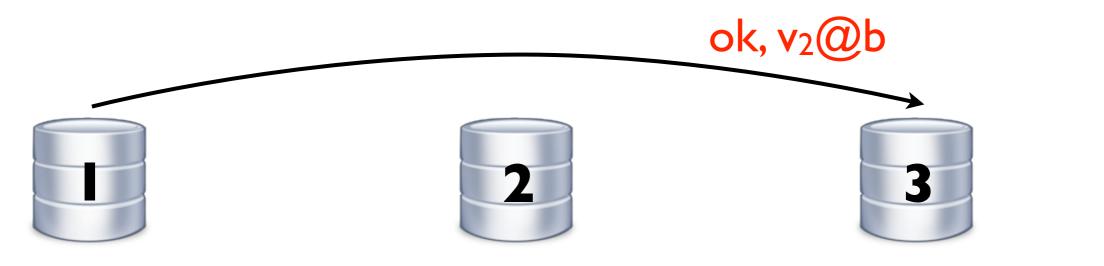


Leader#: 2 Ballot#: b Accepted: v₂@b Reply v₂ to client Leader#: 3 Ballot#: b' > b Accepted: ?



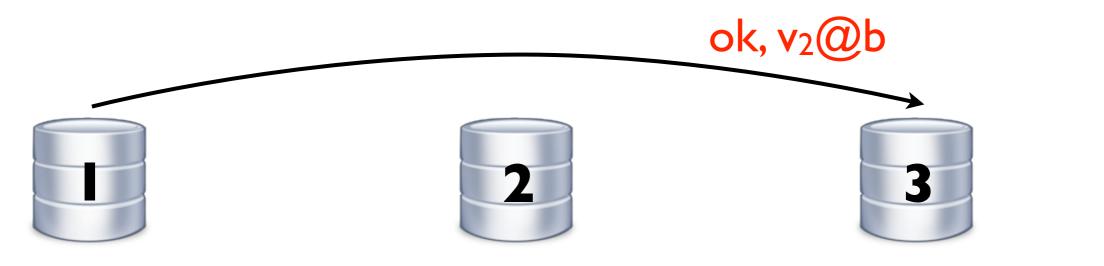
Leader#: 2Leader#: 3Ballot#: bBallot#: b' > bAccepted: $v_2@b$ Accepted: ?Reply v_2 to client

- Phase I: acceptor sends to the prospective leader its value and the ballot it was accepted at
- If some acceptor has accepted a value, the leader proposes the value accepted at the highest ballot number



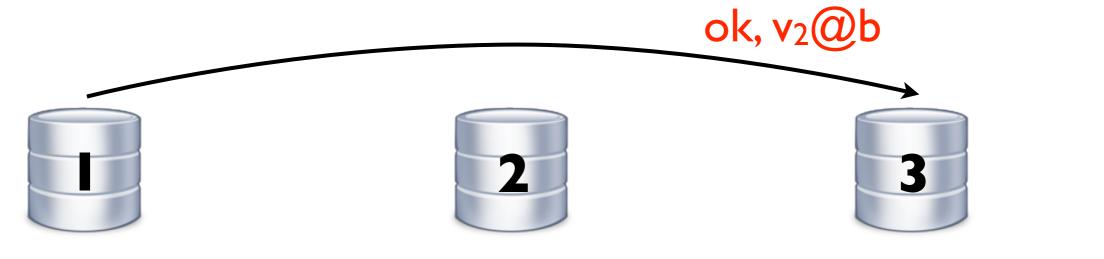
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- Ensures the value chosen will not be changed \implies nodes don't disagree about the chosen value



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Key invariant: If a quorum Q accepted a value v at ballot b, then any leader of a ballot b' > b will also propose v

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- Consider b' > b and assume leader(b'') may only propose v if b < b'' < b'. We prove that leader(b') may only propose v.

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- b' > b
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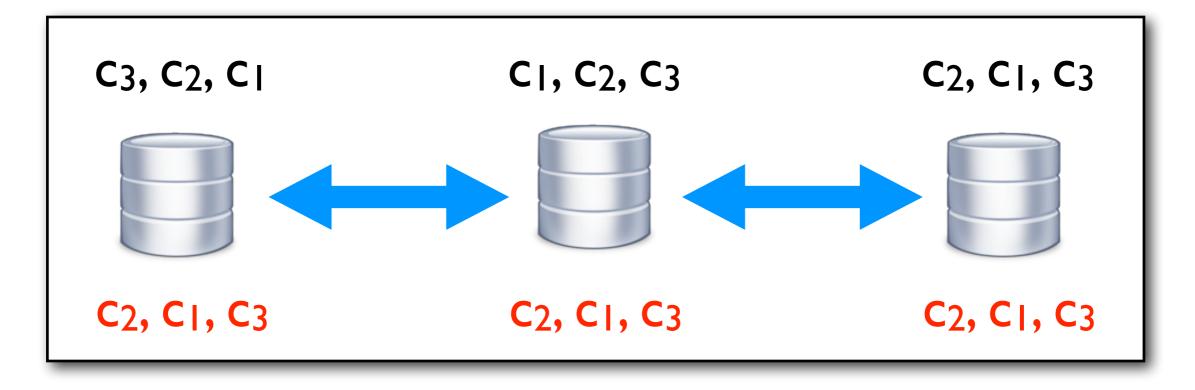
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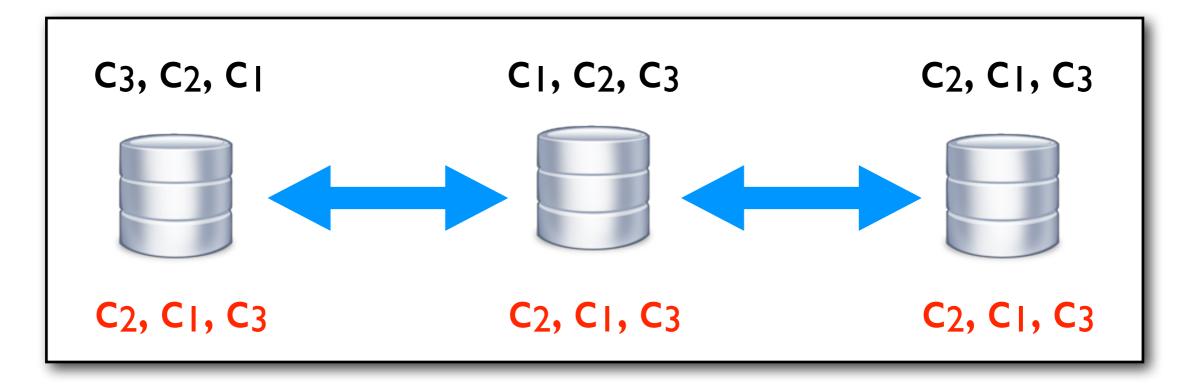
Multi-Paxos

State machine replication requires solving a sequence of consensus instances



Multi-Paxos

State machine replication requires solving a sequence of consensus instances



- Naive solution: execute a separate Paxos instance for each sequence element
- Multi-Paxos: execute Phase I once for multiple sequence elements

Paxos verification

- Lots of work on formally verifying Paxos-like protocols in theorem provers or semi-automatic systems
- Fully automatic verification is an open problem

The end

- Spectrum of data consistency models in distributed systems
- Downsides of weakening consistency can be mitigated by verification techniques and programming abstractions: replicated data types, transactions
- Proving correctness of consistency protocols is a verification challenge