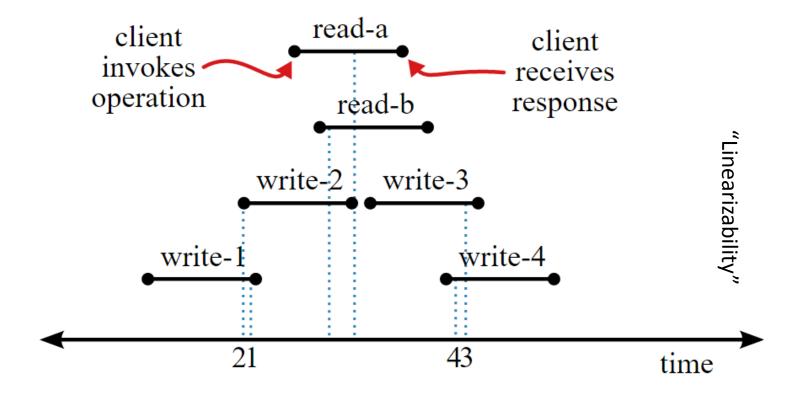
Replication & Consistency



Agenda

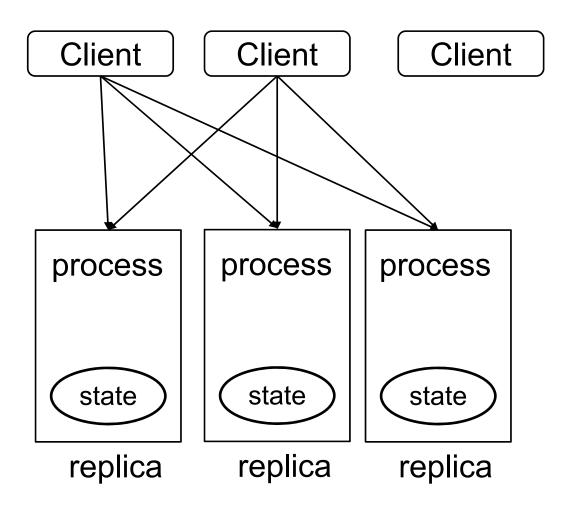
Consistency models

Data-centric consistency

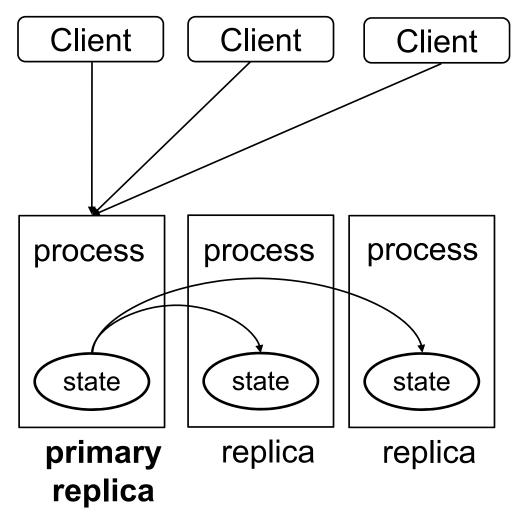
Client-centric consistency

CONSISTENCY MODELS

Active Replication



Passive Replication



Read-Write Conflict

- Alice and Bob buy tickets online, one ticket left
- Alice logs in, sees there is one ticket left, delays purchase
- Bob logs in, sees there is one ticker left, buys the ticket
- Alice resumes, tries to buy ticket now, purchase fails

Read-Write Conflict

Process P1, P2 execute reads and writes concurrently

```
P1 P2
Read(A)
Read(A)
Write(A)
Write(A)
```

Write-Write Conflict

Process P1, P2 execute writes concurrently

P1 P2
Write(A)
Write(B)

Write(B)
Write(A)

- Here, we have P1's version of B and P2's version of A
 - P1's version of A is lost and P2's version of B is lost
- NB: Outcome is not equivalent to any serial execution

Write-Write Conflict Example

(Initially A is 100; Deposit(x): A = A + x)

- Sequential execution to deposit 20:
 - Deposit(10)
 - Deposit(10)
- Implementation of Deposit:
 read A, add 10, store A

- Deposit1(10) || Deposit2(10)
- Deposit1:
 read A {100}, add 10 {110}
- Deposit2: read A {100}
- Deposit1:store A {110}
- Deposit2:
 add 10 {110}, store A {110}

Write-Write Conflict Example

(Initially A is 100; Deposit(x): A = A + x)

has no effect (overwritten by second store).

- Deposit(10)
- Implementation of Deposit:

Update by first deposit is lost (lost update problem).

Deposit1(10) || Deposit2(10)

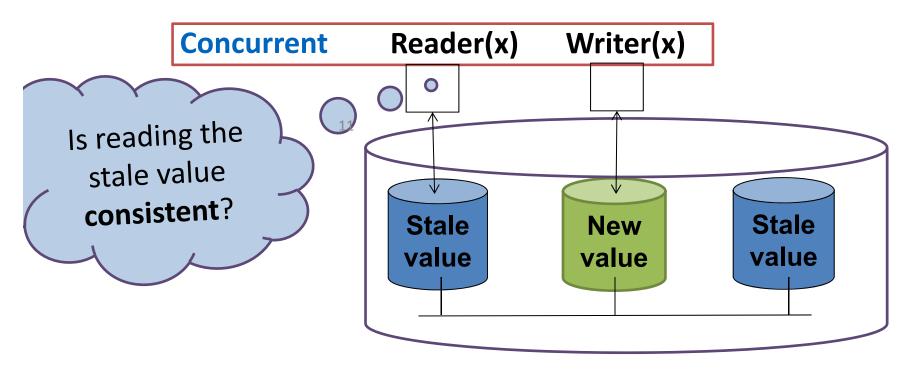
- Deposit1:
 read A {100}, add 10 {110}
- Deposit2: read A {100}
 - Deposit1:

store A {110}

Deposit2:
 add 10 {110}, store A {110}

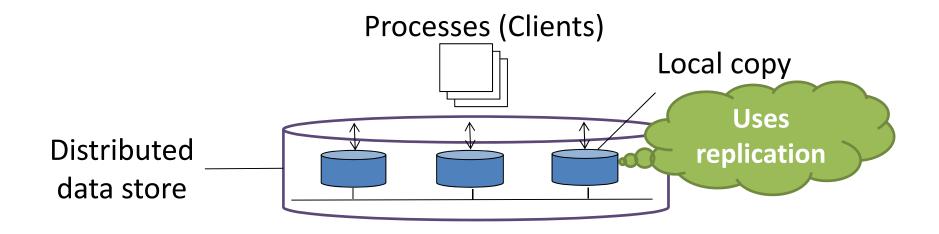
Concurrency and Replication

- All operations must be applied in a specific order to all replicas
- Global ordering is too costly and not scalable (e.g., using consensus, or single primary)
- Solution: Avoid global ordering using weaker consistency requirements suitable for the application



Consistency Models

- Definition (Data-centric Consistency model)
 - A contract between a distributed data store and a set of processes which specifies what the results of concurrent read/write operations are



Distributed data store as synonym for replicas, distributed database, shared memory, shared files, etc.

Data-Centric Consistency Models

- Data-centric consistency models dictate the outcome of concurrent reads and writes (r/w and w/w conflicts):
 - Strict consistency
 - Sequential consistency
 - Linearizable consistency
 - Causal consistency
 - FIFO consistency
 - Weak consistency

Strict Consistency

- Definition: Any read on a data item x returns a value corresponding to the result of the most recent write on x
- Uni-processor systems have traditionally observed strict consistency, ...

but what about multi-processor systems?

 Definition assumes existence of absolute global time for unambiguous determination of "most recent".

Interpretation of Strict Consistency

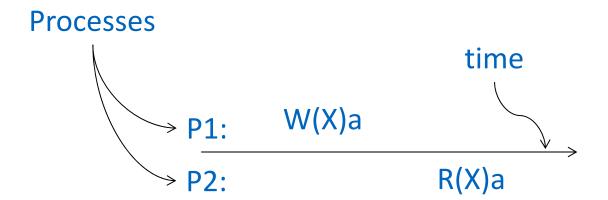
- Under strict consistent, all writes are instantaneously visible
 to all processes and absolute global time order is maintained
- If a replica is updated, ...
 - all subsequent reads, see the new value, no matter how soon after the update the reads are done
 - and no matter which process is doing the reading and where it is located
- Similarly, if a read is done, then it gets the most recent value, no matter how quickly the next write is done

Notation

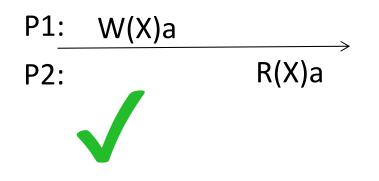
W(X)a: Represents **writing** the value **a** to data **X** (memory)

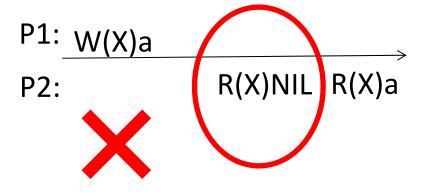
R(X)a: Represents **reading** data **X**, which returns the value **a**

Initial value of X is NIL



Strict Consistency Example

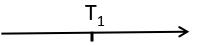




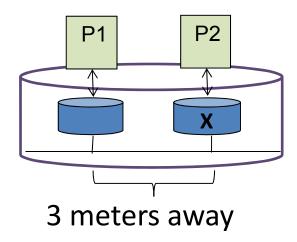
Strict Consistency: Thought Experiment

 To satisfy strict consistency, laws of physics may have to be violated! Obviously, not an option!!!

Example:



@P1: W(X)a at T_1 **@P2:** R(X)a at $T_1 + 1$ ns



To realize strict consistency in this case, W(X)a would have to travel at 10 times the speed of light!

Strict Consistency: The Bad News

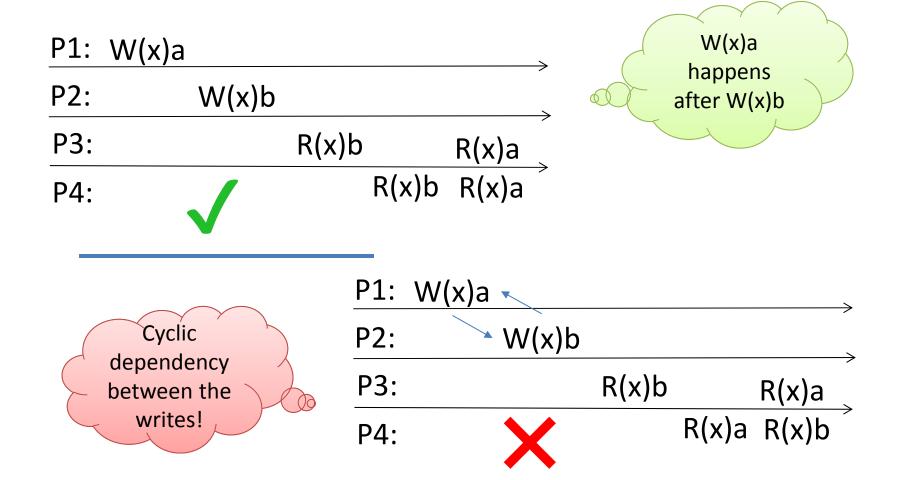
- It is **impossible to perfectly** synchronize clocks
 - How to accurately determine the time of each operation?
- It is **impossible** to instantaneously replicate operations



Definition of Sequential Consistency

- The result of any execution is the same as if the operations by all processes on the data store were executed in some sequential order and ...
- ... the operations of each individual process appear in this sequence in the order specified by its program
- Weaker than strict consistency: logical time instead of physical time

Sequential Consistency Example



Definition of Linearizability

- The result of any execution is the same as if the (read and write) operations because the same as if the (read and write) operations of each individual process appear in this sequence in the order specified by its program.
- In addition, if ts_{OP1}(x) < ts_{OP2}(y), then operation OP1(x) should precede OP2(y) in this sequence
- ts_{OP}(x) denotes the timestamp assigned to operation OP that is performed on data item x, and OP is either read
 (R) or write (W).

Time stamp is modeled as a duration.

Linearizability

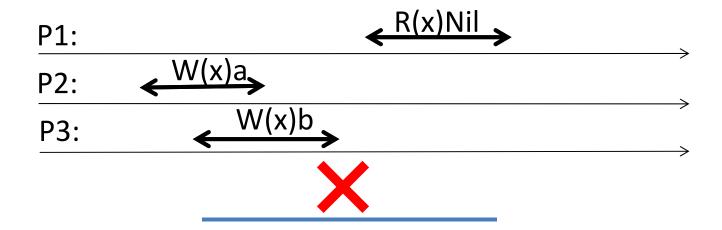
- Like strict consistency, assumes global time, but not absolute global time
- Assumes processes in the system have physical clocks synchronized to within an bounded error (captured by duration timestamp)
- If W(x)b was the **most recent** write operation and there is no other write operation **overlapping** with W(x)b, then any **later** read should return b
- If W(x)a and W(x)b were two most-recent overlapping write operation, then any later read should return either a or b, not something else

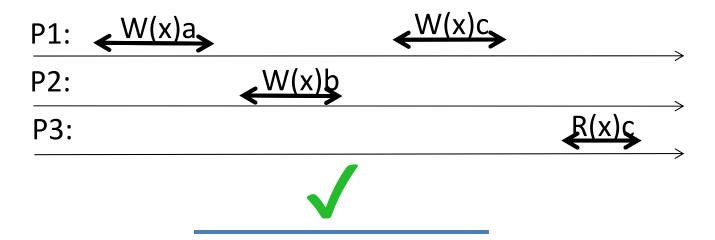
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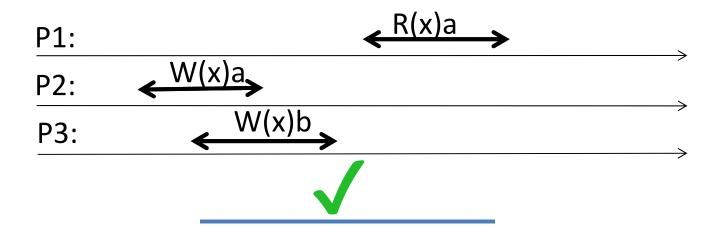


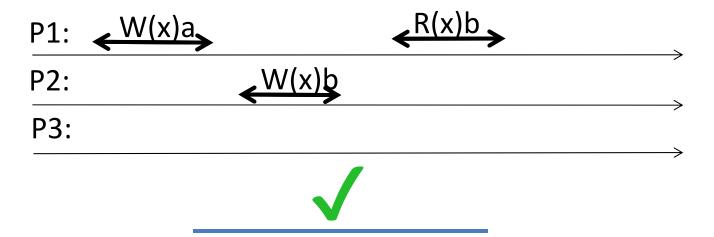
Linearizability

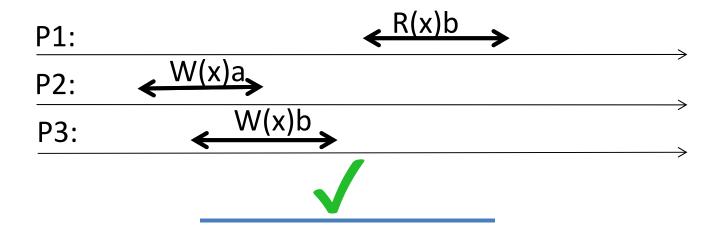
- A linearizable data store is also sequentially consistent
- I.e., linearizability is more restrictive (stronger)
- Difference is the ordering according to a set of synchronized clocks
- Linearizability prevents stale reads, since it guarantees correct reads after a write is completed

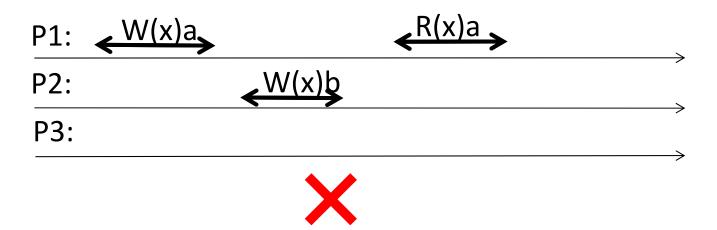


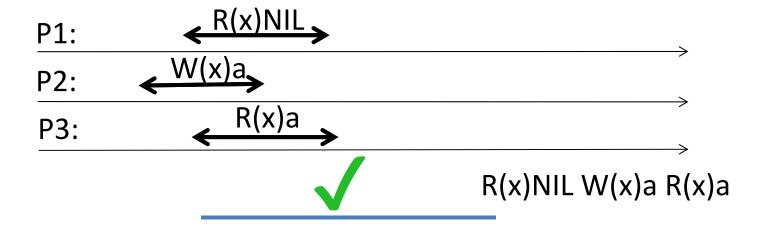


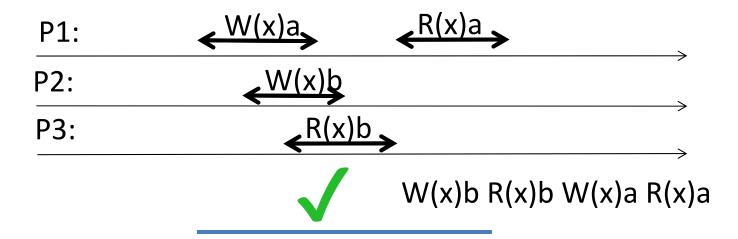


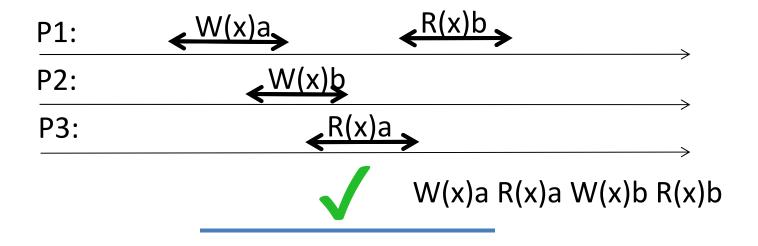


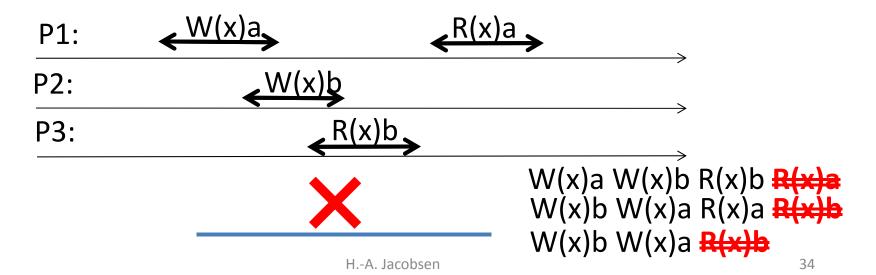












Intuition for Causal Consistency

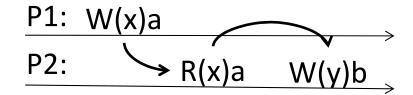
- Weaker than sequential consistency
- Distinguish events that are potentially causally related and those that are not (concurrent events)
- Similar to the happens-before relationship (cf. Lamport clock), but with reads and writes instead of messages
- If Event B is influenced (caused) by an earlier Event A ($A \rightarrow B$), causal consistency requires that every process first sees A then B
- Concurrent events (i.e., writes) may be seen in a different order on different machines

Causal Relationship

- Read followed by write in the same process, the two are causally related
 - **Example**: $R(x)a \rightarrow W(y)b$ (e.g., it may be that y=f(x))
- Read is causally related to write that provided the value read got (across processes)
 - Example: $W(x)a \rightarrow R(x)a$
- Transitivity: if Op1 \rightarrow Op2, Op2 \rightarrow Op3, then Op1 \rightarrow Op3
- Independent writes by two processes on a variable are not causally related (they are concurrent)
 - Example: $W(x)a \mid \mid W(x)b$

"Potentially" Causally Related

• Example: Say, $W_1(x)$, then $R_2(x)$ and $W_2(y)$

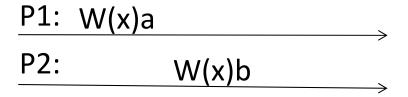


- Reading of x and writing of y are potentially causally related
- Computation of y may have depended on value of x read by P2 (written by P1); e.g., y = f(x)
- On the other hand, y may not have depended on x, yet potential causality still holds in our formalization!

Definition of Causal Consistency

 "Potentially" causally related writes must be seen by all processes in the same order

 Concurrent writes may be seen in a different order by different processes



Causal Consistency Example I

P1: W	V(x)a		W(x)c		
P2:	R(x)a	W(x)b			→
P3:	R(x)a			R(x)c	R(x)b
P4:	R(x)a			R(x)b	R(x)c

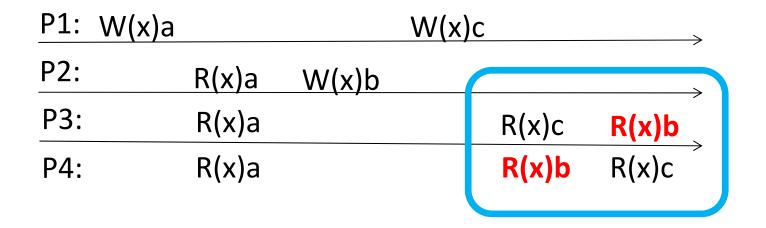
Is this strictly, sequentially or causally consistent?

Causal Consistency Example I

P1:	W(x)a			W(x)c		→
P2:		R(x)a	W(x)b			
P3:		R(x)a			R(x)c	R(x)b
P4:		R(x)a			R(x)b	R(x)c

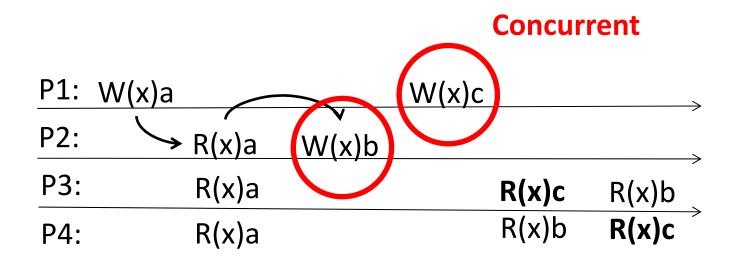
Neither strictly, nor sequentially consistent, but causally consistent.

Causal Consistency Example I



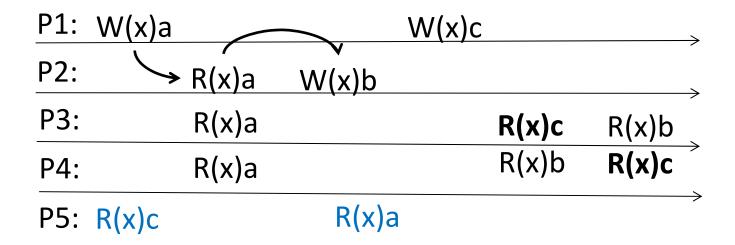
Neither **strictly**, nor **sequentially** consistent, but causally consistent.

Causal Consistency Example I



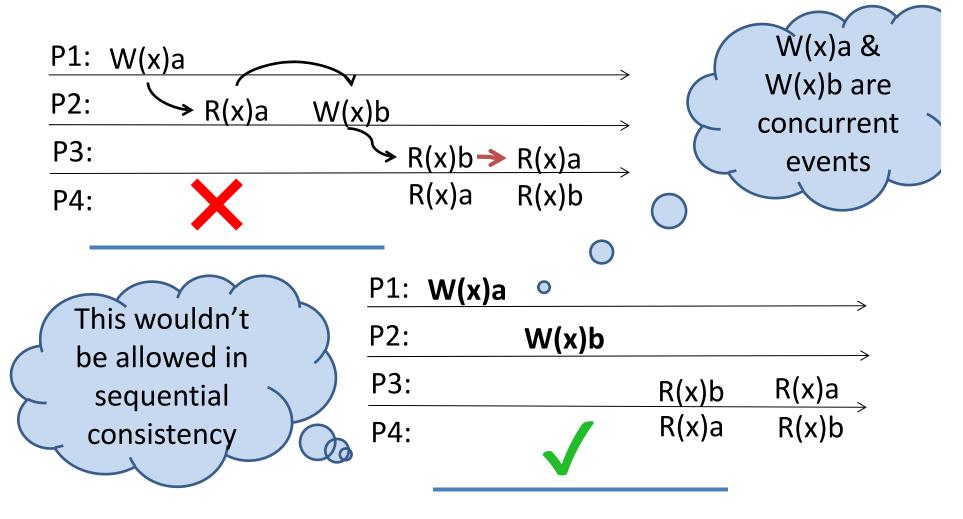
Neither strictly, nor sequentially consistent, but causally consistent!

Causal Consistency Example I



Not causally consistent, W(x)a happens before W(x)c in P1

Causal Consistency Example II



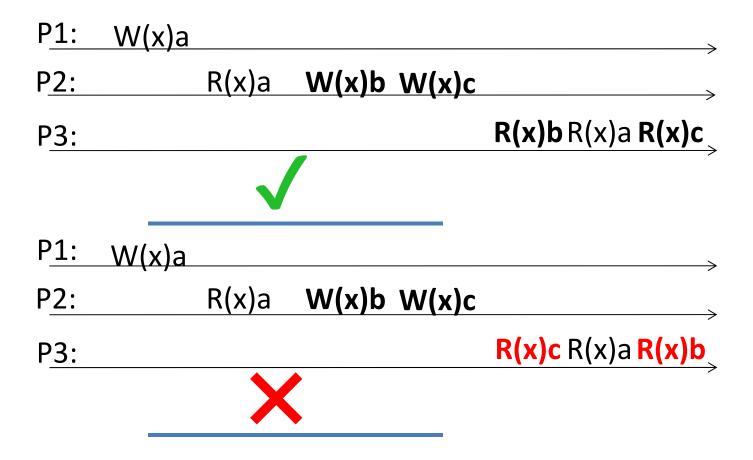
Definition of FIFO Consistency

 Writes by a single process are seen by all other processes in the order in which they were issued

 Writes from different processes may be seen in a different order by different processes

 Easy to maintain: simply send writes in FIFO order from each process to each replica (e.g. using TCP)

FIFO Consistency



Weak Consistency

- Not all processes need to see all writes, let alone in the same order
- Based on synchronization variable S with single operation Synchronize(S)
- A distributed critical section to access shared resources
- Any process can perform read/write operations and synchronize
- Order of operations before synchronization is not consistent
- After a synchronization, all processes see the same outcome
 of operations preceding the synchronization point

Weak Consistency Properties

- Access to synchronization variables are sequentially consistent
- No operation on a synchronization variable is allowed to be performed until all previous writes have been completed at all replicas
- No read or write operation are allowed to be performed until all previous operations to synchronization variables have been performed

Weak Consistency Interpretation

- Process forces the just written value out to all replicas
- Process can be sure to get the most recent value written before it reads
- Model enforces consistency on a group of operations as opposed to individual reads and writes
- Care about the effect of a group of reads and writes

Weak Consister

P2 & P3 have yet to synchronize, no guarantees about values read

P <u>1:</u>	W(x)a	W(x)b	S				
P2:				R(x)a	R(x)b	S	
P3 <u>:</u>				R(x)b	R(x)a	S	
			,	F	P1 propa		es
P1:	W(x)a	W(x)b	S	,00			
P2:				S	R(x)	a	



Weak Consistency

Agreement on value is implementation dependent.

P1 & P2 agree on value a

P <u>1:</u>	W(x)a W(x)b	S	R(x)a	
P2:	W(x)a	S	R(x)a	
P3:			R(x)b R(x)a	S



P1 & P2 agree on value b

P <u>1:</u>	W(x)a W(x)b	5	K(X)	D •••	
P2:	W(x)a	S	R(x)b		
P3:			R(x)b	R(x)a	S



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Weak Consistency

P <u>1:</u>	W(x)a W(x)b	S	R(x)a	9	
P2:	W(x)a	S	R(x)b		
P3:			R(x)b	R(x)a	S



P <u>1:</u>	W(x)a W(x)b	S	R(x)b	
P2:	W(x)a	S	R(x)a	
P3:			R(x)b R(x)a	S



Summary of Consistency Models

Strongest

Consistency	Description
Strict	Absolute time ordering of all shared accesses
Linear- izability	All processes see all shared accesses in the same order . Accesses ordered according to a (non-unique) global timestamp
Sequential	All processes see all shared accesses in the same order . Accesses are not ordered in time
Causal	All processes see causally-related shared accesses in the same order
FIFO	All processes see writes from each other in the order they were issued. Writes from different processes may not always be seen in that order
Weak	Shared data can be counted on to be consistent only after a synchronization is done

Weakest

When to use...?

- Linearizability: Strongest distributed solution, possible with eager replication (synchronous), Hbase, Bigtable
- Sequential: System-wide consistent reads, e.g., everyone sees replies to a post in same order
- Causal: Everyone sees posts before replies
- FIFO: Reading all messages from each friend in order but not across friends
- Weak: Responsibility left to the developer who must explicitly enforce synchronization

CLIENT-CENTRIC CONSISTENCY

Client-Centric Consistency

- So far, the goal was to maintain consistency in presence of concurrent read and write operations
- There are use cases with no (few) concurrent writes, or consistency of write operations are secondary
 - **DNS**: **No write-write conflicts** since there is a single authority updating each domain (disjoint partitions)
 - **Key-value stores**: **Usually no write-write conflicts** since updates partitioned by keys, e.g., Dynamo, Cassandra (also, optimistic concurrency control)
 - WWW: Heavy use of client-side caching, reading stale pages is acceptable in many cases

Client-Centric Consistency

- Client-centric consistency puts the emphasis on maintaining a consistent view for a single process, instead of on the data stored in the system
- Emphasis on read-write conflicts, and we assume write-write conflicts do not exist (i.e., no concurrent writes)
- Client-centric consistency models describe what happens when a single client writes/reads from multiple replicas

Eventual Consistency

- Eventual consistency states all replicas eventually converge when write operations stop
 - E.g., lazy replication using gossiping (cf. replication)
- Very weak form of consistency with no time bound, but highly available (i.e., always return a value, but could be stale)
- Works fine if a client always reads from the same replica...
- ...but gives "weird" results if client reads from multiple replicas:
 - In a mobile scenario
 - During replica failure
- Client-centric consistency models describe what happens when a single client writes/reads from multiple replicas

Notation

Different locations (replicas)

L1: WS(x1) R(x1)

L2: WS(x1;x2) R(x1;x2)

- Reads and writes by one client at different locations
- WS represents a series of write operations at L_i
- E.g., WS(x1) denotes that only op x1 has been written;
 WS(x1;x2) denotes that op x1 and then later op x2 have been written
- R(x1;x2) means the read returns a value formed by operation x1 followed by x2
- W(x1), R(x1) are write to and read from x1

Read-Your-Writes Consistency

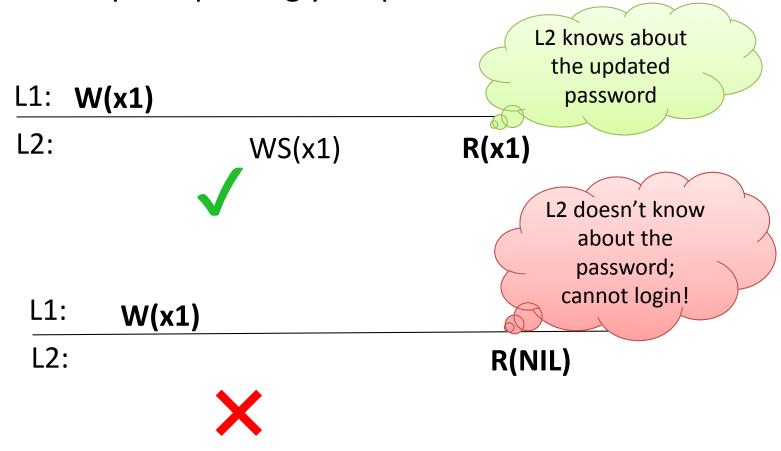
• Informally, ... W ... R ... (by same process!)

 If a read R follows a write then W is included in the set of writes read by R

• A write operation is always completed before a successive read operation by the same process, no matter where the read operation takes place

Read Your Writes

Example: Updating your password on a cluster



Monotonic-Reads Consistency

- ... $WS1(...) R ... WS2(...) R ... (s.t. <math>WS1(...) \subseteq WS2(...)$)
- If a process reads a value formed by a set of operations, WS, any successive read operation of that process will always return a value formed from a superset of WS
- A process always sees more recent data (but not necessarily fresh!)
- If a process reads again from another replica that replica must have already received the relevant older operations

Example: Monotonic Reads

Example: E-mail client

Each read returns the list of emails,

Each write adds one e-mail

L2 knows about all e-R(x1)L1: WS(x1) mails R(x1;x2)_o WS(x1;x2)L2: L2 doesn't know about the e-mail R(x1)WS(x1)x1! R(x2) • L2: WS(x2)

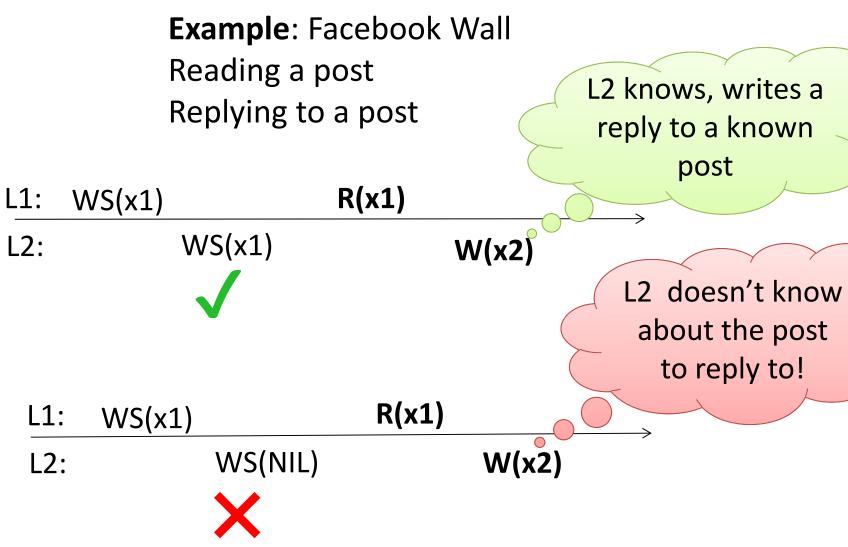
Writes-Follow-Reads Consistency

• ... WS(X1, ..., Xn) **R1** ... WS(X1, ..., Xn) **W2** ...

 If a read R1 precedes a write W2 then at any replica, if W2 is written, it is preceded by WS, the write set read by R1

 Any successive write operation by a process will be performed on a state that is up to date with the value most recently read by that process

Writes Follow Reads



Monotonic-Writes Consistency

• Informally, ... W1 ... W2 ...

 If a write W1 precedes a write W2 then on any replica, W1 must precede W2

 In other words, the writes by the same process are performed in the same order at every replica

Resembles FIFO consistency model

Monotonic Writes

Example: Replicating software code

W(x1) adds a new method to a program

W(x2) calls the new method

L2 knows about the new method

L1: **W(x1)**

L2:

WS(x1)

W(x2)

W(x2) <



L1: **W(x1)**

L2:



know the new method! The code does not

compile

L2 doesn't

Sketch of Mechanisms Realizing CCMs

- Use optimistic concurrency control to verify that the replica has the needed information
- Client keeps track of the read-set and write-set, depending on which consistency model is used
- When processing an operation, the two sets are passed to the replica, who must verify those operations are already processed
- Otherwise, the incoming operation is queued or rejected

Summary: Client-centric consistency

- Eventual consistency can be used when:
 - Few concurrent write occurs
 - Used in Dynamo, Cassandra, Riak
- Eventual consistency provides high availability and scalability
- Client-centric consistency dictates how reads and writes should look from the client's perspective
 - Read your writes
 - Monotonic reads
 - Monotonic writes
 - Writes follow reads