

# Weakening consistency for scalable information systems

#### Arnd Poetzsch-Heffter

Based on work of Annette Bieniusa together with Marc Shapiro, Marek Zawirski (INRIA & LIP6) Nuno Preguiça, Sérgio Duarte (UNL) Carlos Baquero (U. Minho)

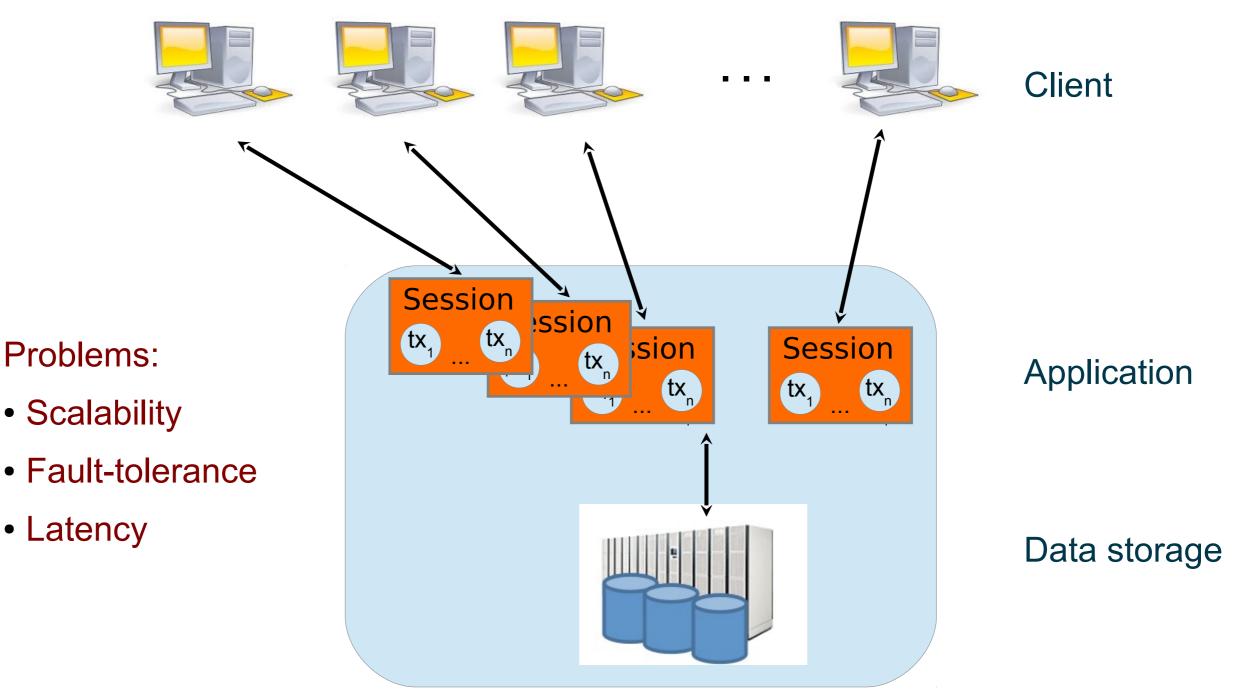
1

#### Overview

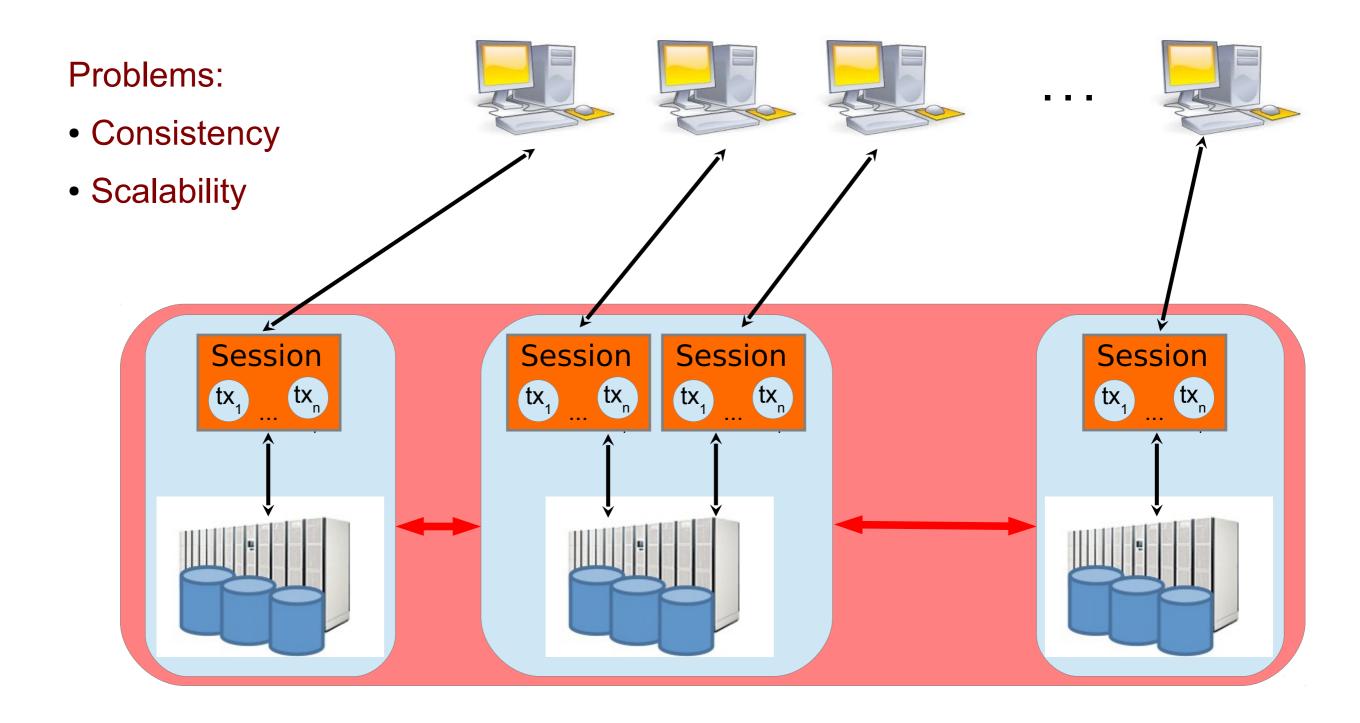
- Strong vs. eventual consistency
- Conflict-free replication
- Realizing future information systems

# Strong VS. eventual consistency

# Centralized information systems



# Strongly consistent, distributed information systems



#### Discussion

Objectives:

- $\rightarrow$  fault tolerance  $\rightarrow$  redundancy
- $\rightarrow$  low latencies  $\rightarrow$  distribution
- ▹ simple to program → strong-consistency

CAP (Brewer `00; Gilbert & Lynch `06) Strongly-Consistent  $\cap$  Available  $\cap$  Partition-Tolerant =  $\emptyset$ 

#### Way out: Give up strong consistency

#### Eventual consistency

#### Basic ideas:

- Clients can live with weaker forms of consistency
- Update each replica independently
  - transport changes to other replicas
  - replay or merge
- Guaranteed delivery:
  - eventually, all replicas receive all updates
  - hopefully they converge... (otherwise: conflicts)
  - but order of updates differs!

### Using eventual consistency

Different approaches:

- Application-specific vs. general approaches
- Conflict resolution:
  - manual
  - automatic
  - no conflicts
- Convergence:
  - ad hoc / programmed
  - guaranteed

# Conflict-free replication

### Strong eventual Consistency

Update local + propagate:

- Update is durable
- Broadcast
- No synchronization

*No* conflict:

 Unique outcome of updates (& propagations)

# Assumptions for strong eventual consistency

Eventual delivery:

Every update eventually executes at all replicas.

Termination:

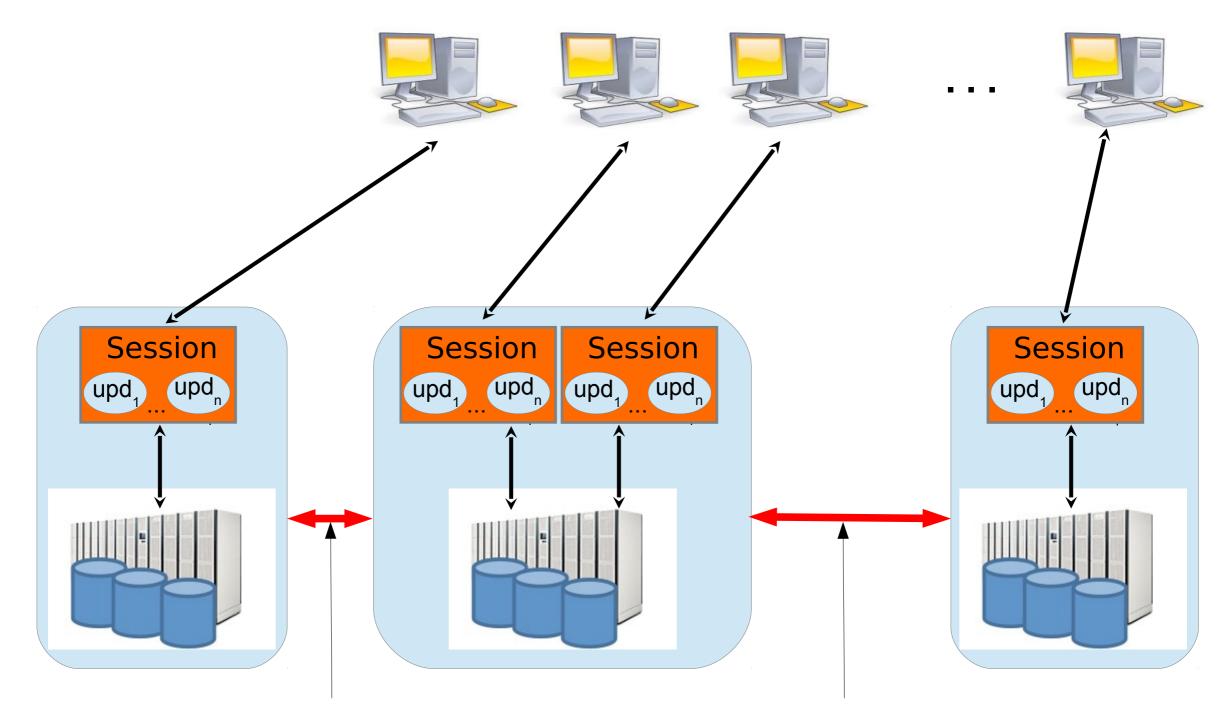
Every update terminates.

#### **Strong** convergence:

Correct replicas that have executed the same updates have equivalent state.

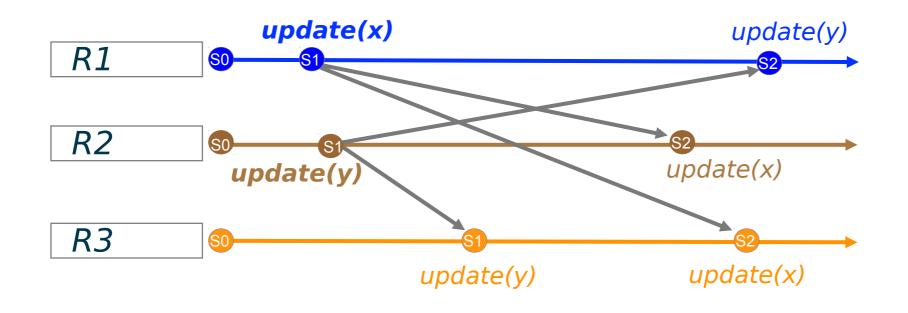
Deterministic, context-independent outcome to concurrent updates

# Conflict-free replicated data types (CRDTs)



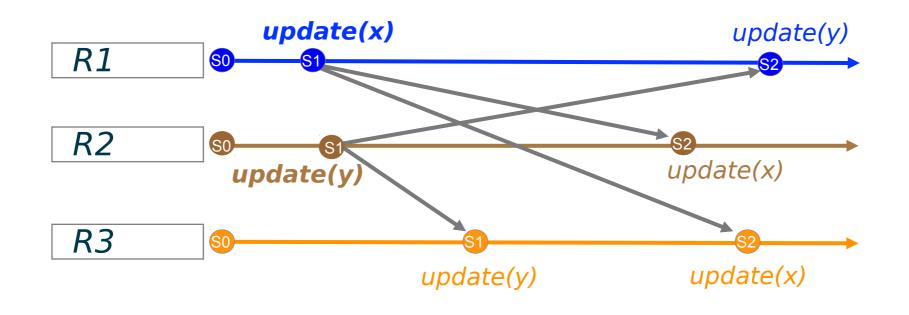
#### **Asynchronous propagation**

### **Operation-based updates**



- Small messages, no information duplication
- Uses causal broadcast
  - Vector clock counts messages received / node
  - Size of vector clock ~ number of replicas
- Consensus not required

### **Operation-based CRDTs**

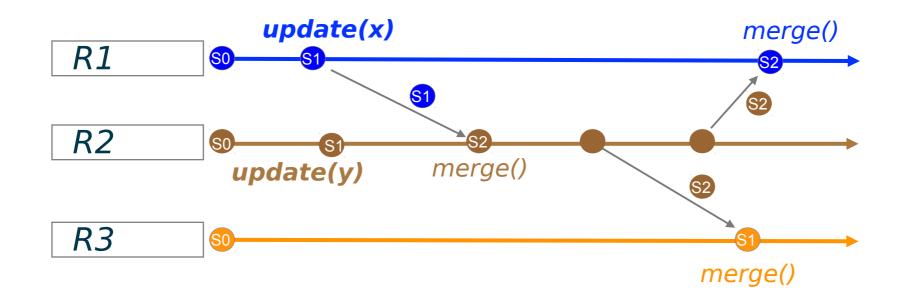


- Example: Counter with *incr* and *decr*
- All replicas have equivalent state in the end
- Sufficient condition:
  - Reliable causal delivery of vector clocks
  - Concurrent operations commute

# Operation-based specification

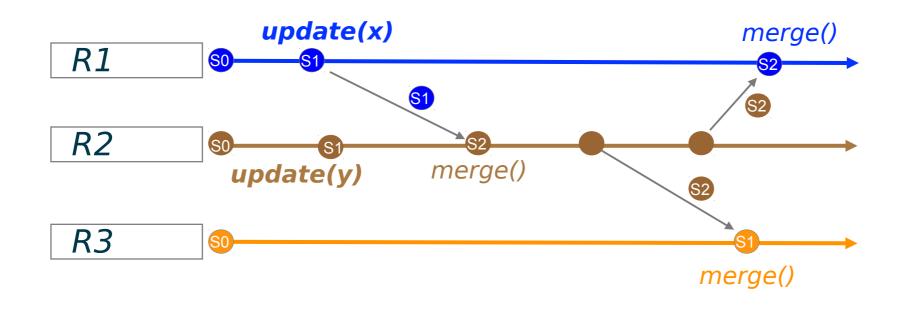
payload Payload type; instantiated at all replicas initial Initial value query Source-local operation (arguments) : returns pre Precondition let Execute at source, synchronously, no side effects update Global update (arguments) : returns prepare (arguments) : returns pre Precondition at source let 1st phase: synchronous, at source, no side effects (arguments passed downstream) effect pre Precondition against downstream state 2nd phase, asynchronous, side-effects to downstream state

### State-based / data shipping



- Epidemic propagation
- Eventual delivery
- Consensus not required
- Inefficient for large payload
- Convergence

#### State-based CRDTs

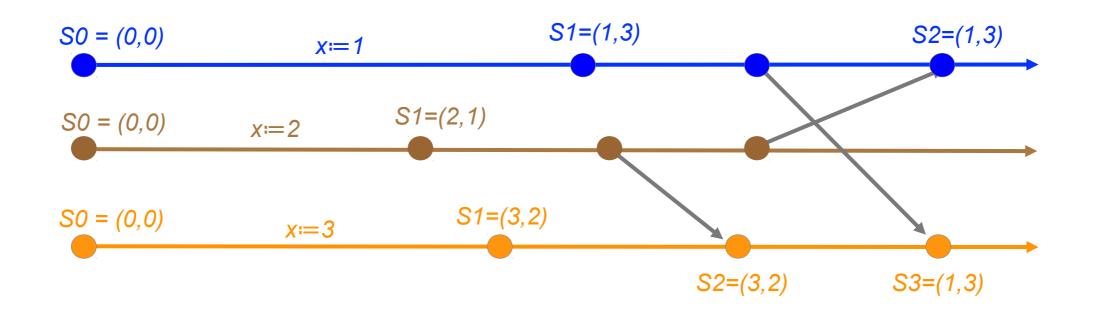


- All replicas have equivalent state in the end
- Sufficient condition: monotonic semi-lattice
  - partial order
  - monotonic
  - merge computes LUB
  - merge eventually delivered

### State-based specification

payload Payload type; instantiated at all replicas initial *Initial value* query Query (arguments) : returns pre Precondition let Evaluate synchronously, no side effects update Source-local operation (arguments) : returns pre Precondition let Evaluate at source, synchronously Side-effects at source to execute synchronously compare (value1, value2) : boolean bIs value 1 < value 2 in semilattice? merge (value1, value2) : payload mergedValue LUB merge of value1 and value2, at any replica

#### Last-writer-wins register



Payload

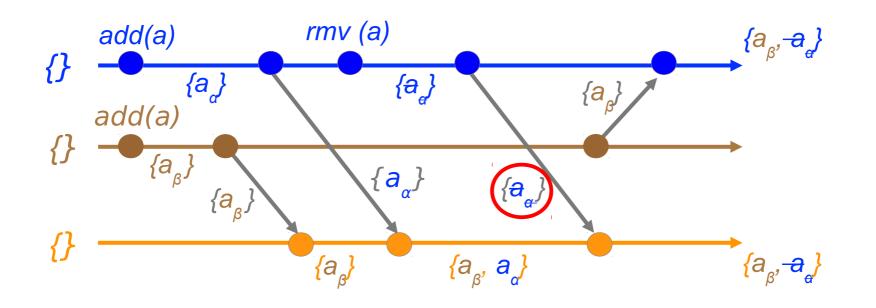
 $S \stackrel{\text{\tiny def}}{=}$  (value v, timestamp ts)

Update  $S \bullet [x \coloneqq v] \stackrel{\text{def}}{=} (v, \text{ now}())$ 

Merge  $S \bullet merge(S') \stackrel{\text{def}}{=} S.ts < S'.ts ? S' : S$ 

Compare  $S \leq S' \stackrel{\text{def}}{=} S.ts \leq S'.ts$ 

### Observed-remove Set



Payload

Update

 $S \stackrel{\text{\tiny def}}{=} (A = \{(e, uid), \ldots\}, R = \{(e', uid'), \ldots\})$ 

 $S \bullet add(e) \stackrel{\text{\tiny def}}{=} (A \cup \{(e, uid)\}, R)$  $S \bullet rmv(e) \stackrel{\text{\tiny def}}{=} (A \setminus T, R \cup T) \text{ with } T = \{(e, \_) \in A\}$ 

Lookup

 $S \bullet lookup(e) \stackrel{\text{\tiny def}}{=} e \in A$ 

Merge

- $S \bullet merge(S') \stackrel{\text{\tiny def}}{=} (A \setminus R' \cup A' \setminus R, R \cup R')$
- Compare  $S \leq S' \stackrel{\text{\tiny def}}{=} A \cup R \subseteq A' \cup R' \land R \subseteq R'$

### Further examples of CRDTs

#### Register

- Last-Writer Wins
- Multi-Value

Set

- Grow-Only
- 2P
- Observed-Remove
  Map

#### Counter

- Unlimited
- Non-negative

Graph

- Directed
- Monotonic DAG
- Edit graph

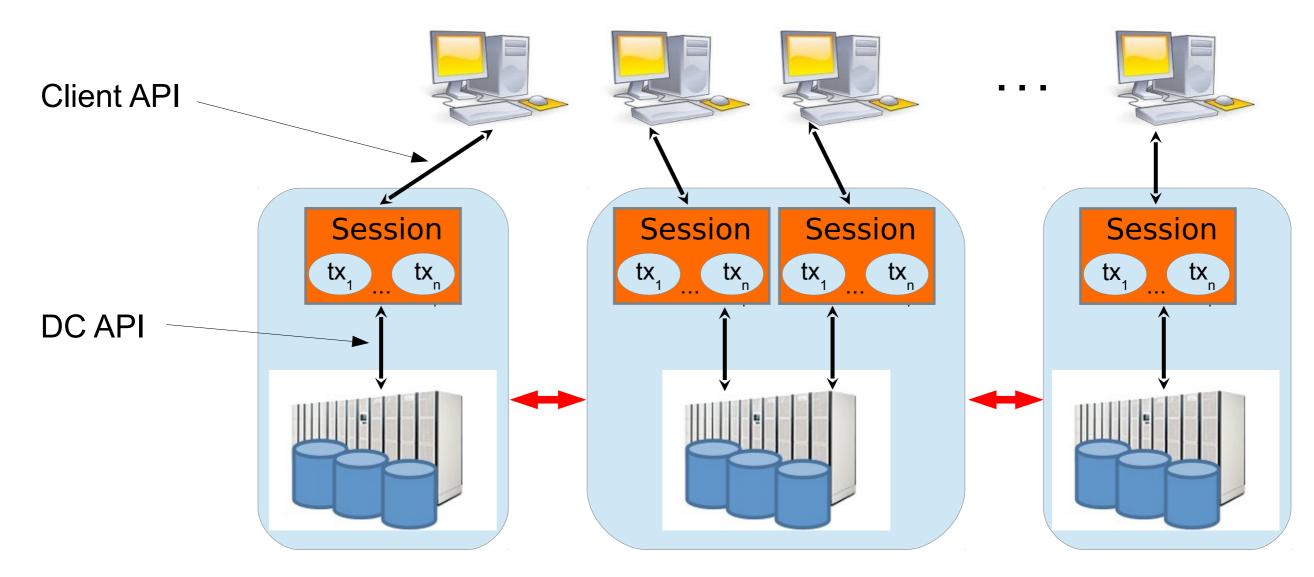
Sequence

### Summary: CRDT

- Concurrent updates have deterministic outcome
- Sufficient conditions:
  - State-based: epidemic, monotonic semi-lattice
  - Op-based: causal, concurrent  $\Rightarrow$  commute
- CRDTs
  - don't lose updates
  - converge eventually
  - have durable updates, no rollbacks
  - support unlimited (crash-recovery) failures

# Realizing future information systems

#### Programming model



#### **Central questions:**

- What is the application-independent API of data store?
- How can CRDTs be combined to realize client API?
- What is needed in addition to CRDTs?

### Further challenges

- More complex architectures:
  - client state
  - DC hopping
- Global state guarantees:
  - support of reservations
  - stable preconditions
- Transactions
- Verification techniques
- Using CRDTs for concurrent programming