

A Generic Specification Framework for Weakly Consistent Replicated Data Types

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Abstract—Burckhardt et al. proposed a formal specification framework for eventually consistent replicated data types, denoted (vis, ar) , based on the notions of visibility and arbitration relations. However, being specific to eventually consistent systems, this framework has two limitations. First, it does not cover non-convergent consistency models since arbitration ar is a total order over events. Second, it does not cover the consistency models in which each event is required to be aware of the return values of some events that are visible to it when justifying its return value. These limitations make the (vis, ar) framework not generic enough to specify and reason about important weak consistency models such as Causal Memory and PRAM. In this paper, we extend this framework to a more generic one called (vis, ar, V) for weakly consistent replicated data types. To specify non-convergent consistency models as well, we relax the arbitration relation ar to be a partial order. To overcome the second limitation, we allow to specify for each event e , a subset $V(e)$ of its visible set whose return values cannot be ignored when justifying the return value of e . To make it practically feasible, we provide candidates for the visibility and arbitration relations and the V function. By combining candidates for these three components, we are able to specify not only existing consistency models but also new ones that are reasonable and promising for practical usefulness. We then show how to specify consistency models in our framework, and provide three case studies.

Index Terms—Replicated Data Types, Consistency Models, Sequential Consistency, Causal Consistency, Pipelined Consistency

1 INTRODUCTION

GEORGRAPHICALLY distributed systems often replicate data at multiple sites to achieve high availability and low latency, even under network partitions [1], [2]. According to the CAP theorem [3], [4] and the PACELC tradeoff [5], these systems often sacrifice strong consistency and choose to implement *weakly consistent replicated data types*.

Eventual consistency is one of the most widely used weak consistency models in distributed systems [6], [7]. It guarantees that “if clients stop issuing updates, the replicas will eventually reach a consistent state.” [7]. For example, to allow replicas to respond to user operations immediately, collaborative text editing systems [8], [9] usually implement an *eventually consistent* replicated list object modelling the shared document. It requires the final lists at all replicas to be identical after executing the same set of user operations [8]. In principle, eventual consistency has two aspects:

- 1) At the *strong* aspect, it requires eventual *convergence* among replicas and thus the clients¹ will eventually obtain the same view of replica states.

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1. Following [2], we use clients, sessions, and processes interchangeably. We also use program order and session order interchangeably.

- 2) At the *weak* aspect, it imposes no restrictions on the intermediate states and thus the clients may obtain (temporarily) inconsistent views of replica states.

Burckhardt et al. [1], [2] proposed a formal specification framework for eventually consistent replicated data types. It is based on the *visibility* (denoted vis) and *arbitration* (denoted ar) relations over events of a history, and we call it the (vis, ar) framework.

- The visibility relation is an acyclic relation that accounts for the relative timing of events. Intuitively, if an event e_1 is visible to another event e_2 , it means that the effect of e_1 is visible to the client performing e_2 before e_2 is invoked. For example, e_2 may be a query returning the value written by update e_1 [10]. We call two events *concurrent* if they are invisible to each other.
- The arbitration relation is a *total order* over all events of a history. It indicates how the system resolves the conflicts due to concurrent events. For example, such a total order can be achieved using timestamps [11].

Being specific to eventually consistent systems, the (vis, ar) framework is not general enough to cover some important weak consistency models such as PRAM [12] and Causal Memory [13]. Specifically, we identify two limitations of the (vis, ar) framework as follows, corresponding to the two aspects of eventual consistency mentioned above:

- 1) *Convergence vs. Non-Convergence*. The total ordering requirement of ar over all events aims to ensure eventual *convergence* of the clients’ views of replica states. Therefore, the (vis, ar) framework does not cover the consistency models that do not enforce convergence.
- 2) *Awareness vs. Unawareness*. In the (vis, ar) framework, the return value of an event e is justified by the set $vis^{-1}(e)$ of events visible to e (arranged in the order

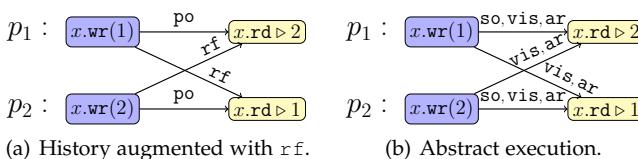


Fig. 1: Motivating history for convergence and non-convergence: It satisfies *CM*, but it is non-convergent. We denote the operation of writing value v to x by $x.wr(v) \triangleright \perp$ (where \perp indicates that it returns no value and is omitted in figures) and the operation of reading v from x by $x.rd \triangleright v$. We write $x.rd \triangleright \perp$ to emphasize that the return value has been ignored.

ar), while *ignoring* all their return values. Therefore, the *(vis, ar)* framework does not cover the consistency models in which an event is required to be *aware* of the return values of some events that are visible to it.

The feature of unawareness is useful for wait-free asynchronous message-passing distributed systems where operations should be completed without waiting for any other process. Unawareness allows such systems to execute operations in a speculative order with the possibility of rolling back and reordering previously executed operations if needed [14], [15], [16]. For example, Bayou [6] makes both local operations and remote operations received immediately visible, even if there are prior operations in the arbitration that may arrive in the future. When an operation o is received, all operations arbitrated after o but have already been applied are undone and then reapplied after applying o [16].

We illustrate both limitations of the *(vis, ar)* framework with Causal Memory (*CM*) [13], [14] with respect to read/write registers, and motivate our first contribution of this paper which is a generalization of the *(vis, ar)* framework. Intuitively, *CM* ensures that processes agree on the relative ordering of operations that are *causally related* [13]. The *causality order* over operations is defined as the transitive closure of the union of program order *po* and the read-from relation *rf* which informally associates each read with a unique write from which it reads the value. A history satisfies *CM* (w.r.t. read/write registers) if for each process, the set of all operations on this process and all write operations on other processes can be arranged into an operation sequence which preserves the causality order such that each read reads the value from the most recently preceding write in this sequence on the same register.

Example 1. Consider the history in Fig. 1(a) consisting of two processes p_1 and p_2 which read from and write to a shared register x . This history satisfies *CM*: the witness operation sequences for p_1 and p_2 are $\langle x.wr(1) \ x.wr(2) \ x.rd \triangleright 2 \rangle$ and $\langle x.wr(2) \ x.wr(1) \ x.rd \triangleright 1 \rangle$, respectively. However, it is non-convergent: processes p_1 and p_2 cannot agree with a total order *ar* over $x.wr(1)$ and $x.wr(2)$. Particularly, it does *not* satisfy the convergent variant of causal consistency called *WCCv* [2], [14] defined in the *(vis, ar)* framework; see Fig. 1(b) and Definition 17.

This example demonstrates that the *(vis, ar)* framework

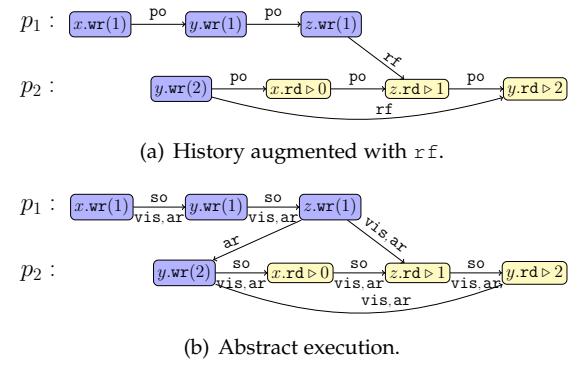


Fig. 2: Motivating history for awareness and unawareness. It violates *CM*. The arrows for relations implied by transitivity are not drawn.

is insufficient for specifying histories that adhere to non-convergent consistency models such as *CM*. To specify non-convergent consistency models as well, we can relax the arbitration relation *ar* to be a partial order.

Example 2. Consider the history in Fig. 2(a). It does not satisfy *CM*. In any operation sequence for process p_2 , $x.wr(1)$ must be placed after $x.rd \triangleright 0$ and $z.wr(1)$ must be placed before $z.rd \triangleright 1$. As a consequence, $y.wr(1)$ must be placed between $y.wr(2)$ and $y.rd \triangleright 2$. However, such sequences cannot be valid; in particular, the return value of $y.rd \triangleright 2$ is not justified.

However, if the *return values* of the operations that causally precede $y.rd \triangleright 2$ can be ignored, $y.rd \triangleright 2$ can be justified by the operation sequence $\langle x.wr(1) \ y.wr(1) \ z.wr(1) \ y.wr(2) \ x.rd \triangleright 0 \ z.rd \triangleright 1 \ y.rd \triangleright 2 \rangle$. Similarly, $z.rd \triangleright 1$ and $x.rd \triangleright 0$ (0 is the initial value of x) can be justified by $\langle x.wr(1) \ y.wr(1) \ z.wr(1) \ y.wr(2) \ x.rd \triangleright 0 \ z.rd \triangleright 1 \rangle$ and $\langle y.wr(2) \ x.rd \triangleright 0 \rangle$, respectively. Actually, this history in Fig. 2(a) satisfies *WCCv* [2], [14] defined in the *(vis, ar)* framework in which return values are ignored; see Fig. 2(b) and Definition 17.

This example demonstrates that the *(vis, ar)* framework is insufficient for specifying histories that adhere to consistency models that demand awareness. To allow an event e to be aware of the return values of some or all events in $\text{vis}^{-1}(e)$, we introduce a function V defined on events. Specifically, $V(e)$ for event e is a subset of $\text{vis}^{-1}(e)$ whose return values cannot be ignored when justifying the return value of e .

Our First Contribution. In this paper, we extend the *(vis, ar)* specification framework for eventually consistent replicated data types into a generic one called *(vis, ar, V)* for weakly consistent replicated data types (Section 3). On the one hand, by relaxing *ar* to be a partial order, the *(vis, ar, V)* framework is able to cover non-convergent consistency models such as PRAM [12] and Causal Memory [13]. On the other hand, by introducing the V function for events, the *(vis, ar, V)* framework is able to cover the consistency models in which each event is required to be aware of the return values of some or all events that are visible to it. To make it practically feasible, we provide common candidates for the three components of this generic frame-

work, including the *vis* and *ar* relations and the *V* function.

Our Second Contribution. By combining candidates for each component, we are able to specify various consistency models, including both existing ones and new ones, in the (vis, ar, V) framework, including variants of sequential consistency (Section 4), causal consistency (Section 5), and pipelined consistency (Section 7). In particular, we find that three variants of sequential consistency are equivalent. We also provide case studies to demonstrate the usefulness of our framework. Specifically, we show that the causal consistency protocol of MongoDB in the failure-free sharded cluster deployment satisfies Causal Memory Convergence, a new causal consistency variant discovered in our framework (Section 6). For pipelined consistency, we show that although the failure-free client-server implementation of GSP (Global Sequence Protocol) satisfies Weak Causal Convergence, it does not satisfy Pipelined Convergence, another pipelined consistency variant discovered in our framework (Section 8). We also show that RA-Linearizability, a specification recently proposed for conflict-free replicated data types, implies Weak Pipelined Convergence, but not Pipelined Convergence (Section 9).

This paper is an extended version of the SRDS conference paper [17], with four new Sections 4, 7, 8, and 9.

2 PRELIMINARIES

In this section, we review the formal specification framework called (vis, ar) for eventually consistent replicated data types proposed by Burckhardt et al. [1], [2].

2.1 Relations and Orderings

A binary relation R over a given set A is a subset of $A \times A$, i.e., $R \subseteq A \times A$. For $a, b \in A$, we use $(a, b) \in R$ and $a \xrightarrow{R} b$ interchangeably. The inverse relation of R is denoted by R^{-1} , i.e., $(a, b) \in R \iff (b, a) \in R^{-1}$. We use $R^{-1}(b)$ to denote the set $\{a \in A \mid (a, b) \in R\}$.

Given two binary relations R and S over set A , we define the composition of them as $R; S = \{(a, c) \mid \exists b \in A : a \xrightarrow{R} b \xrightarrow{S} c\}$. For $n \in \mathbb{N}^+$, R^n denotes the n -ary composition $R; R; \dots; R$. The transitive closure of R is $R^+ \triangleq \bigcup_{n \geq 1} R^n$. For some subset $A' \subseteq A$, the restriction of R to A' is $\bar{R}|_{A'} \triangleq R \cap (A' \times A)$. If $f : A \rightarrow B$ is a function (also a relation) from A to B , the restriction of f to $A' \subseteq A$ is $f|_{A'} \triangleq f \cap (A' \times B) = \{(a, f(a)) \mid a \in A'\}$.

A relation R is *natural* if $\forall x \in A : |R^{-1}(x)| < \infty$. A (strict) partial order is an irreflexive and transitive relation. A total order is a relation which is a partial order and total. For $A' \subseteq A$, $\text{to}(R, A')$ asserts that R is a total order over A' .

2.2 Abstract Data Types

We consider a replicated database storing *objects* of some abstract data types.

Definition 1. An *abstract data type* $\tau \in \text{Type}$ is a pair $\tau = (Op, Val)$ such that

- Op is the set of *operations* supported by τ ;
- Val is the set of values allowed by τ . We assume $\perp \in Val$ to indicate that some operations may return no value.

Definition 2. The *sequential semantics* of a type τ is a function $\text{eval}_\tau : Op^* \times Op \rightarrow Val$ that, given a sequence of operations S and an operation o , determines the return value $\text{eval}_\tau(S, o) \in Val$ for o when o is performed after S .

Examples 3, 4, and 5 contain formal definitions for the intuitive sequential semantics of read/write registers reg , key-value stores kvs , and queues fq , respectively.

Example 3. An integer read/write register reg supports two operations: $\text{wr}(v)$ writes value $v \in \mathbb{Z}$ to the register and rd reads the value from it. A rd operation returns the value of the last preceding wr , or the initial value 0 if there are no prior writes. Formally, for any operation sequence S ,

$$\begin{aligned} \text{eval}_{\text{reg}}(S, \text{wr}(v)) &= \perp, \\ \text{eval}_{\text{reg}}(S, \text{rd}) &= v, \text{ if } \text{wr}(0) S = S_1 \text{ wr}(v) S_2 \\ &\quad \text{and } S_2 \text{ contains no wr operations.} \end{aligned}$$

Example 4. A key-value store kvs supports two operations: $\text{PUT}(k, v)$ writes value v to key k and $\text{GET}(k)$ reads value (which may be the initial value 0) from key k . Formally, for any operation sequence S ,

$$\begin{aligned} \text{eval}_{\text{kvs}}(S, \text{PUT}(k, v)) &= \perp, \\ \text{eval}_{\text{kvs}}(S, \text{GET}(k)) &= v, \text{ if } \text{PUT}(k, 0) S = S_1 \text{ PUT}(k, v) S_2 \\ &\quad \text{and } S_2 \text{ contains no PUT operations on } k. \end{aligned}$$

Example 5. An integer FIFO queue fq supports two operations. $\text{enq}(v)$ adds value $v \in \mathbb{N}$ to the tail of the queue. deq removes and returns the element v at the head of the queue; if the queue is empty, we let $v = \perp$. Formally, for any operation sequence S ,

$$\begin{aligned} \text{eval}_{\text{fq}}(S, \text{enq}(v)) &= \perp, \\ \text{eval}_{\text{fq}}(S, \text{deq}) &= \begin{cases} \perp, & \text{if } S \text{ is empty} \\ v, & \text{if } S = S_1 \text{ enq}(v) S_2 \\ & \text{and } \text{eval}_{\text{fq}}(S_1, \text{deq}) = \perp \\ & \text{and } S_2 \text{ contains no deq.} \end{cases} \end{aligned}$$

2.3 Histories

Clients interact with the replicated database by performing operations on objects. We use a *history* to record such interactions in a computation.

Definition 3. A *history* is a tuple $H = (E, op, rval, so)$ such that

- E is the set of all *events* of operations invoked by clients in a single computation;
- $op : E \rightarrow Op$ describes the operation of an event;
- $rval : E \rightarrow Val$ describes the value returned by the operation $op(e)$ of an event e ;
- $so \subseteq E \times E$ is a partial order over E , called the *session order*. It relates operations within a session in the order they were invoked by clients.

We lift op to sets of events by defining $op(F) = \{op(e) \mid e \in F\}$ for $F \subseteq E$. See Figs. 1(b) and 2(b) for examples of histories (for now ignore the relations *vis* and *ar*).

2.4 Abstract Executions

To justify the return values of all events in a history, we need to know how these events are related to each other. Following [1], [2], this is captured declaratively by the *visibility* and *arbitration* relations.

Definition 4. An *abstract execution* is a triple $A = ((E, \text{op}, \text{rval}, \text{so}), \text{vis}, \text{ar})$ such that

- $(E, \text{op}, \text{rval}, \text{so})$ is a history;
- Visibility $\text{vis} \subseteq E \times E$ is an acyclic and natural relation;
- Arbitration $\text{ar} \subseteq E \times E$ is a total order.

Figs. 1(b) and 2(b) show examples of abstract executions. In both abstract executions, the visibility relations vis correspond to the read-from rf relations. The arbitration relation $z.\text{wr}(1) \xrightarrow{\text{ar}} y.\text{wr}(2)$ in Fig. 2(b) enforces an order between these two concurrent operations.

2.5 Consistency Models

Definition 5. A *consistency model* is a set of consistency predicates on abstract executions.

A consistency predicate on an abstract execution A is a statement that is true or false depending on A [2]. We write $A \models P$ if the consistency predicate P is true on the abstract execution A .

Definition 6. An *abstract execution* A satisfies consistency model $\mathcal{C} = \{P_1, \dots, P_n\}$, denoted $A \models \mathcal{C}$, if each consistency predicate in \mathcal{C} is true on A .

We are concerned with histories that satisfy some consistency model.

Definition 7. A *history* H satisfies consistency model $\mathcal{C} = \{P_1, \dots, P_n\}$, denoted $H \models \mathcal{C}$, if it can be extended to an abstract execution that satisfies \mathcal{C} . Formally, $H \models \mathcal{C} \iff \exists \text{vis}, \text{ar}. (H, \text{vis}, \text{ar}) \models \mathcal{C}$.

2.6 Return Value Consistency

A common consistency predicate is the consistency of return values defined for a given data type τ , denoted $\text{RVAL}(\tau)$. $\text{RVAL}(\tau)$ requires that the return value $\text{rval}(e)$ of an event e in an abstract execution A should agree with the result computed by applying the sequential semantics of data type τ to the operation sequence based on $\text{ctxt}_A(e)$, which is obtained by arranging the events in $\text{vis}^{-1}(e)$ according to ar , and the operation $\text{op}(e)$ [2]. Here $\text{ctxt}_A(e)$ is called the *operation context* of e in A . It is the restriction of A to the set $\text{vis}^{-1}(e)$ of events visible to e .

Definition 8. The *operation context* of an event $e \in E$ in an abstract execution $A = ((E, \text{op}, \text{rval}, \text{so}), \text{vis}, \text{ar})$ is defined as $\text{ctxt}_A(e) \triangleq A|_{\text{vis}^{-1}(e), \text{op}, \text{vis}, \text{ar}}$.

Definition 9. For a data type τ , its *return value consistency* predicate on an abstract execution A is $\text{RVAL}(\tau) \triangleq \forall e \in E. \text{rval}(e) = \text{eval}_\tau(\text{ctxt}_A(e), \text{op}(e))$.

Note that $\text{ctxt}_A(e)$ ignores rval in the original history. Consequently, when justifying the return value of event e , it is *not* required for eval to be consistent with the return values of the events visible to e . As shown in Example 2, the abstract execution in Fig. 2(b) satisfies $\text{RVAL}(\text{reg})$.

3 A GENERIC SPECIFICATION FRAMEWORK FOR WEAKLY CONSISTENT REPLICATED DATA TYPES

In this section, we present our generic $(\text{vis}, \text{ar}, V)$ specification framework for weakly consistent replicated data types. It not only covers more existing consistency models than the (vis, ar) framework does, but also helps to discover new ones. We first define the $(\text{vis}, \text{ar}, V)$ framework in a general way. Then, we provide candidates for the visibility and arbitration relations and the V function.

3.1 The $(\text{vis}, \text{ar}, V)$ Specification Framework

The $(\text{vis}, \text{ar}, V)$ framework generalizes (vis, ar) in two ways, overcoming the two limitations of (vis, ar) :

- 1) *Convergence vs. Non-Convergence*. The arbitration relation ar in $(\text{vis}, \text{ar}, V)$ is a partial order over events. Therefore, $(\text{vis}, \text{ar}, V)$ is able to cover classic non-convergent consistency models such as PRAM [12] and Causal Memory [13].
- 2) *Awareness vs. Unawareness*. In $(\text{vis}, \text{ar}, V)$, we allow to specify for each event e , a subset $V(e)$ of $\text{vis}^{-1}(e)$ whose return values cannot be ignored when justifying the return value of e . Therefore, $(\text{vis}, \text{ar}, V)$ is able to cover the consistency models in which each event is required to be aware of the return values of some or all events that are visible to it.

To cover non-convergent consistency models, we reformulate the definition of abstract executions by relaxing ar to be a partial order.

Definition 10. An *abstract execution* in the $(\text{vis}, \text{ar}, V)$ framework is a triple $A = ((E, \text{op}, \text{rval}, \text{so}), \text{vis}, \text{ar})$ such that

- $(E, \text{op}, \text{rval}, \text{so})$ is a history;
- Visibility $\text{vis} \subseteq E \times E$ is an acyclic and natural relation;
- Arbitration $\text{ar} \subseteq E \times E$ is a partial order.

To be aware of return values is a bit more involved. We first reformulate operation context by selectively unhiding the return values of some or all visible events.

Definition 11. Let $A = ((E, \text{op}, \text{rval}, \text{so}), \text{vis}, \text{ar})$ be an abstract execution in the $(\text{vis}, \text{ar}, V)$ framework. The *operation context* of $e \in E$ in A is defined as

$$\text{ctxt}_A(e, V) \triangleq A|_{\text{vis}^{-1}(e), \text{op}, \text{rval}|_{V(e)}, \text{vis}, \text{ar}},$$

where $V : E \rightarrow 2^E$ specifies a subset of $\text{vis}^{-1}(e)$. We let Ctx be the set of all operation contexts, ranged over by \mathbb{C} . We use V_C to select the set $V(e)$ in \mathbb{C} .

Accordingly, the RVAL consistency predicate should be adapted to use $\text{ctxt}_A(e, V)$.

Definition 12. For a data type τ , its *return value consistency* predicate on an abstract execution A is

$$\text{RVAL}(\tau, V) \triangleq \forall e \in E. \text{rval}(e) \in \text{eval}_\tau(\text{ctxt}_A(e, V), \text{op}(e)).$$

In the definition above, we need to redefine the sequential semantics of τ , i.e., the eval_τ function. On the one hand, since ar is a partial order, there may be a set of operation sequences over $\text{vis}^{-1}(e)$ to evaluate. Thus, we regard an operation context as a *set of serializations*, which are linear extensions of ar over $\text{vis}^{-1}(e)$. This is why we use ' \in '

TABLE 1: Recipes for vis , ar , and V in $(\text{vis}, \text{ar}, V)$ Framework.

vis	$\emptyset \subseteq \text{vis}$
	$\text{so} \subseteq \text{vis}$ (Read Your Writes [18])
	$\text{vis}; \text{so} \subseteq \text{vis}$ (Monotonic Reads [18])
	$\text{so}; \text{vis} \subseteq \text{vis}$ (Monotonic Writes [18])
	$\text{so}; \text{vis}; \text{so} \subseteq \text{vis}$
	$\text{vis}; \text{so}; \text{vis} \subseteq \text{vis}$ (Writes Follow Reads [18])
ar	$(\text{so} \cup \text{vis})^+ \subseteq \text{vis}$
	$\emptyset \subseteq \text{ar}$
	$\text{so} \subseteq \text{ar}$
	$\text{vis} \subseteq \text{ar}$
	$\text{vis}; \text{so} \subseteq \text{ar}$
$V(e)$	$\text{to}(\text{ar}, E)$
	$V(e) = \emptyset$
	$V(e) = \text{so}^{-1}(e) \cap \text{vis}^{-1}(e)$
	$V(e) = \text{vis}^{-1}(e)$

instead of '=' in the definition of $\text{RVAL}(\tau, V)$. On the other hand, besides justifying the return value $\text{rval}(e)$ of event e , eval_τ is required to preserve the unhidden return values specified by $\text{rval}|_{V(e)}$ in a given serialization obtained from $\text{ctxt}_A(e, V)$. Such serializations are considered *valid*.

Definition 13. The *sequential semantics* of a type $\tau \in \text{Type}$ is a function $\text{eval}_\tau : \text{Ctxt} \times \text{Op} \rightarrow 2^{\text{Val}}$ that, given an operation context $\mathbb{C} \in \text{Ctxt}$ and an operation $o \in \text{Op}$, determines the possible return values $\text{eval}_\tau(\mathbb{C}, o) \subseteq \text{Val}$ for o when o is performed in the context \mathbb{C} . Specifically, $\text{eval}_\tau(\mathbb{C}, o)$ is computed as follows²:

$$\begin{aligned} \text{eval}_\tau(\mathbb{C}, o) = \{v \in \text{Val} \mid \exists S \in \mathbb{C} : \\ (\text{eval}_\tau(\text{op}(S), o) = v \wedge \\ \forall e \in V_\mathbb{C} : \text{rval}(e) = \text{eval}_\tau(\text{op}(S_{\prec e}), \text{op}(e)))\}, \end{aligned}$$

where $S_{\prec e}$ is the prefix of S before e . Given a serialization $S \in \mathbb{C}$, the first conjunction computes the return value for o when o is performed after operation sequence $\text{op}(S)$, and the second one ensures that S is valid by checking the unhidden return values w.r.t. corresponding prefixes of S .

Example 6. We argue that the history in Fig. 1 satisfies CM in the $(\text{vis}, \text{ar}, V)$ framework. (CM is formally defined in Section 5.) We choose an ar which does not arbitrate between $x.\text{wr}(1)$ and $x.\text{wr}(2)$. It is easy to check that the resulting abstract execution in Fig. 1(b) satisfies $\text{RVAL}(\text{reg})$.

Example 7. We argue that the history in Fig. 2 does not satisfy CM in the $(\text{vis}, \text{ar}, V)$ framework. Consider the abstract execution in Fig. 2(b), obtained by augmenting the history with vis and ar as in Example 2. Besides $\text{vis} \subseteq \text{ar}$, CM requires eval_{reg} to preserve the unhidden return values of the visible events in $V(e) = \text{so}^{-1}(e)$. By similar argument to that in Example 2, there is no way of justifying $y.\text{rd} \triangleright 2$ while preserving the return values of $x.\text{rd} \triangleright 0$ and $z.\text{rd} \triangleright 1$.

3.2 Recipes for the $(\text{vis}, \text{ar}, V)$ Framework

The $(\text{vis}, \text{ar}, V)$ framework is parametric w.r.t. three components, i.e., the vis and ar relations and the V function. By combining different consistency predicates on them, we can specify various consistency models in this framework. To make it practically feasible, we provide a number of candidates for each component, as summarized in Table 1.

2. Note that we overload eval_τ with different type signatures.

3.2.1 Recipe for the Visibility Relation

We identify a number of common consistency predicates on vis in roughly the order they induce larger and larger visible sets consisting of the events visible to some event.

- The weakest one does not enforce an event to observe any particular set of events. Formally, $\emptyset \subseteq \text{vis}$.
- The most basic ingredient for visibility is the session order so . The predicate $\text{so} \subseteq \text{vis}$ requires each event to see all the previous events *in the same session*.
- To allow an event to see the events *in different sessions* as well, it is necessary to compose so with vis in some ways. There are four basic kinds of compositions, namely $\text{vis}; \text{so} \subseteq \text{vis}$, $\text{so}; \text{vis} \subseteq \text{vis}$, $\text{so}; \text{vis}; \text{so} \subseteq \text{vis}$, and $\text{vis}; \text{so}; \text{vis} \subseteq \text{vis}$.
- To further allow an event to see the events in different sessions through an arbitrarily long chain of compositions of vis and so , we rely on the transitive closure over vis and so . That is, we have $(\text{so} \cup \text{vis})^+ \subseteq \text{vis}$, where $\text{hb} \triangleq (\text{so} \cup \text{vis})^+$ is the well-known *happens-before order* [2], [1] first proposed by Lamport [11]. Note that $\text{hb} \subseteq \text{vis}$ implies that vis is transitive.

3.2.2 Recipe for the Arbitration Relation

When justifying the return value of event e with respect to the sequential semantics eval , we rely on ar to resolve conflicts caused by concurrent events in the set $\text{vis}^{-1}(e)$ of events visible to e . In the $(\text{vis}, \text{ar}, V)$ framework, ar is a partial order. In the following, we identify a number of common consistency predicate on ar in roughly the order they are able to resolve more and more conflicts.

- The weakest one does not impose any constraints on how the conflicts should be resolved. Formally, $\emptyset \subseteq \text{ar}$.
- A slightly stronger arbitration orders the events in $\text{vis}^{-1}(e)$ (for some event e) according to the session order [1]. Formally, $\text{so} \subseteq \text{ar}$. Note that so may be a *proper* subset of vis .
- To resolve all conflicts (using ar) among visible events to an event, we require $\text{vis} \subseteq \text{ar}$.
- $(\text{vis}; \text{so}) \subseteq \text{ar}$ orders an event after other ones previously observed in the same session [1].
- Finally, convergent consistency models often require ar to be a total order over E , denoted $\text{to}(\text{ar}, E)$, as in the (vis, ar) framework [1], [2].

3.2.3 Recipe for the V Function

By definition, $V(e)$ is a subset of $\text{vis}^{-1}(e)$. We identify three common consistency predicates on $V(e)$:

- The strongest one requires e to be aware of the return values of *all* events visible to it. Formally, $V(e) = \text{vis}^{-1}(e)$;
- Consistency models like Causal Memory [13] allow e to ignore the return values of its visible events *in different sessions* while being aware of those in its own session. Formally, $V(e) = \text{so}^{-1}(e) \cap \text{vis}^{-1}(e)$;
- The weakest one allows e to ignore the return values of any visible events, as in the (vis, ar) framework. Formally, $V(e) = \emptyset$.

TABLE 2: Consistency variants specified in the (vis, ar, V) specification framework. (New variants are marked with [*].)

Consistency Models		Alternative Names	vis	ar	$V(e)$	Case Studies	
Sequential Consistency Variants	SC	SEQUENTIALCONSISTENCY (Def. 21 [10])	vis = ar	to(ar, E)	$V(e) = \emptyset$	—	
					$V(e) = so^{-1}(e)$	—	
					$V(e) = vis^{-1}(e)$	—	
Causal Consistency Variants	WCC (Def. 8 [14]) CM ([13], [15]) SCC	CC (Def. 4.2 [15]) CC (Def. 9 [14]) [*] (Section 5.4)	hb \subseteq vis	vis \subseteq ar	$V(e) = \emptyset$	—	
					$V(e) = so^{-1}(e)$	—	
					$V(e) = vis^{-1}(e)$	—	
	WCCv CMv SCCv	CCv (Def. 4.5 [15], Def. 12 [14]) CAUSALCONSISTENCY (Section 5.2 [2]) CAUSALITY (Def. 26 [10]) [*] (Section 5.6) [*] (Section 5.7)		vis \subseteq ar \wedge to(ar, E)	$V(e) = \emptyset$	—	
					$V(e) = so^{-1}(e)$	MongoDB (Section 6) The causal consistency protocol of MongoDB satisfies CMv.	
					$V(e) = vis^{-1}(e)$	—	
Pipelined Consistency Variants	WPC PC SPC	[*] (Section 7.2) PRAM [12] (Def. 2.2 [19]) [*] (Section 7.4)	so \subseteq vis	vis \subseteq ar	$V(e) = \emptyset$	—	
					$V(e) = so^{-1}(e)$	—	
					$V(e) = vis^{-1}(e)$	—	
	WPCv PCv SPCv	SUC (Def. 9 [20]) [*] (Section 7.6) [*] (Section 7.7)		vis \subseteq ar \wedge to(ar, E)	$V(e) = \emptyset$	RA-Linearizability (Section 9) RA-Linearizability implies WPCv, but not PCv.	
					$V(e) = so^{-1}(e)$	GSP (Section 8) The implementation of GSP satisfies WCCv, but not PCv.	
					$V(e) = vis^{-1}(e)$	—	

4 SEQUENTIAL CONSISTENCY: MODELS

Sequential consistency [21] is considered a strong consistency model. It requires that all operations appear to have executed in some sequential order consistent with the session order. We consider three variants of sequential consistency, namely SC_{\emptyset} , SC_{so} , and SC_{vis} (which is the original Sequential Consistency (SC) in [21], [2]). They require $to(ar, E)$ to enforce a total order over all events and $vis = ar$ to capture that all events in $ar^{-1}(e)$ are visible to event e . They differ in the function V . Specifically, SC_{\emptyset} , SC_{so} , and SC require $V(e)$ to be \emptyset , $so^{-1}(e)$, and $vis^{-1}(e)$, respectively. Interestingly, we find that SC_{\emptyset} , SC_{so} , and SC are equivalent.

Theorem 1. Given a history H , we have

$$H \models SC_{\emptyset} \iff H \models SC_{so} \iff H \models SC.$$

Proof. The “ \Leftarrow ” direction holds trivially. Now suppose that $H \models SC_{\emptyset}$. Therefore, there exists a visibility relation vis and an arbitration relation ar such that $vis = ar$ and the return value of each event e in H can be justified by the prefix of ar , i.e., $ar^{-1}(e) = vis^{-1}(e)$, while ignoring their return values. We show that these vis and ar relations also satisfy the properties required by SC_{so} and SC . The proof proceeds by induction on the events in the total ar order.

- (Base Case) The first event, denoted e_1 , in ar can be justified by vis and ar in SC_{so} and SC , since $so^{-1}(e_1) = vis^{-1}(e_1) = \emptyset$.
- (Induction Hypothesis) Assume that for each of the first $(k - 1)$ events e in ar , the return value of e can be justified by $ar^{-1}(e)$, while respecting the return values of events in $so^{-1}(e)$ (resp. $vis^{-1}(e)$) as required by SC_{so} (resp. SC).
- (Inductive Step) Consider the k -th event in ar , denoted e_k . On the one hand, since $H \models SC_{\emptyset}$, the return value of e_k can be justified by $ar^{-1}(e_k)$. On the other hand, by induction hypothesis, the return values of the events in $so^{-1}(e_k)$ (resp. $vis^{-1}(e_k)$) can be justified using these vis and ar as required by SC_{so} (resp. SC).

5 CAUSAL CONSISTENCY: MODELS

5.1 Overview

Causal consistency is one of the most widely used weak consistency models in distributed systems [13], [22]. The key notion is the *happens-before* order hb over events [11], [2]. Intuitively, causal consistency ensures that if an event e_1 happens before event e_2 , then all sessions must observe e_1 before e_2 . However, concurrent events may be observed in different orders by different sessions.

In the literature, there are several causal consistency variants with subtle differences [14], [2], [15]. In this section, we consider six variants that we call Weak Causal Consistency (WCC), Weak Causal Convergence (WCCv), Causal Memory (CM), Causal Memory Convergence (CMv), Strong Causal Consistency (SCC), and Strong Causal Convergence (SCCv), as defined in Table 2. Note that these variants may have different names in related work, as summarized in Table 2. Among these variants, CMv, SCC, and SCCv are new variants discovered in our framework.

In terms of visibility, all causal consistency variants require $hb = (so \cup vis)^+ \subseteq vis$ to capture the happens-before order over events. In terms of arbitration, they all require $vis \subseteq ar$ to enforce the happens-before order over events in $vis^{-1}(e)$ when justifying the return value of the event e in its operation context. However, they may differ in two aspects:³

- How large is the function V for specifying the subset of visible events whose return values must be respected? Note that in causal consistency models, $so^{-1}(e) \subseteq vis^{-1}(e)$ for any event e (since $so \subseteq vis$). Thus, the candidate for the V function $V(e) = so^{-1}(e) \cap vis^{-1}(e)$ is equivalent to $V(e) = so^{-1}(e)$.
- How strong is the arbitration relation ar for resolving conflicts? We distinguish between two cases according to whether ar is a total order or not, given $vis \subseteq ar$.
 - 3. To exclude the trivial implementations in which replicas update objects locally without communicating with each other, we may additionally require the *eventual visibility* (EV) predicate [2], [14]. EV ensures that an event can be invisible to at most finitely many other events.

Fig. 3 shows the relative strength of these variants and Fig. 4 gives examples on objects of FIFO queues.

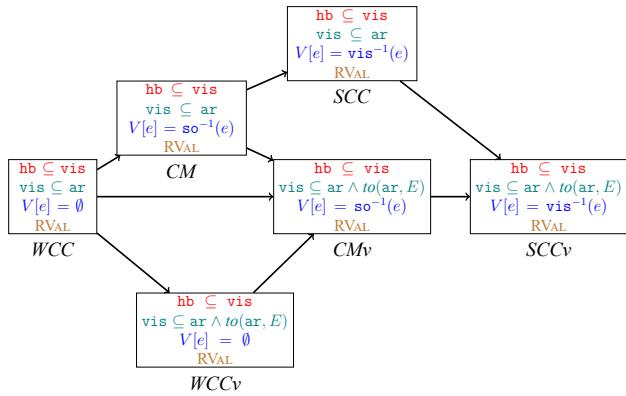


Fig. 3: Causal consistency variants and their relative strength. An arrow from a consistency model \mathcal{C}_1 to another one \mathcal{C}_2 denotes that \mathcal{C}_2 is stronger than \mathcal{C}_1 .

5.2 Weak Causal Consistency

Weak causal consistency (WCC), recently proposed in [14], is the weakest variant of causal consistency we consider. Informally speaking, an abstract execution satisfies WCC as long as the return value $rval(e)$ of each event e can be justified by some serialization of its visible set $vis^{-1}(e)$, while ignoring their return values. More specifically, in terms of arbitration, WCC makes no extra restrictions on convergence and thus concurrent events may be observed in different orders by different sessions. In terms of $V(e)$, WCC allows each event e to ignore all the return values of its visible events.

Definition 14 (Weak Causal Consistency).

$$WCC \triangleq (hb \subseteq vis) \wedge (vis \subseteq ar) \wedge (V(e) = \emptyset) \wedge RVAL.$$

Example 8. The history of Fig. 4(a) does not satisfy WCC. Intuitively, event $q.enq(2)$ (resp. $p.enq(2)$) must be visible to event $q.deq \triangleright 2$ (resp. $p.deq \triangleright 2$). By transitivity of vis , event $p.enq(1)$ is visible to event $p.enq(2)$. Since $vis \subseteq ar$, it is impossible for the third session to dequeue 2 before 1 from the FIFO queue q .

The history of Fig. 4(b) satisfies WCC. For example, the return value of $b : deq \triangleright 2$ can be justified by the serialization $\langle enq(1) enq(2) a : deq \triangleright b : deq \triangleright 2 \rangle$.⁴ Note that such a justification is not required by WCC to be consistent with the return value of $a : deq \triangleright 2$. Similarly, the return value of $b : deq \triangleright 1$ can be justified by the serialization $\langle enq(2) enq(1) a : deq \triangleright b : deq \triangleright 1 \rangle$.

5.3 Causal Memory

Causal memory (CM) was originally defined by Ahamed et al. [13] on read/write registers. Recently, Perrin et al. [14] extended it to arbitrary replicated data types. CM is stronger than WCC in that when justifying the return value $rval(e)$ of each event e , CM takes into account not only the *operation invocations* of the set $vis^{-1}(e)$ of events visible to e as in

4. For clarity, we include the event b to be justified in the serialization.

WCC but also *the return values* of the set $so^{-1}(e)$ of events that precede e in the same session. In other words, compared to WCC, CM requires that each session is consistent with respect to the previous return values provided [15].

Definition 15 (Causal Memory).

$$CM \triangleq (hb \subseteq vis) \wedge (vis \subseteq ar) \wedge (V(e) = so^{-1}(e)) \wedge RVAL.$$

Example 9. Although the history of Fig. 4(b) satisfies WCC, it does not satisfy CM. Specifically, being aware of the return value of $a : deq \triangleright 2$, $b : deq \triangleright 2$ is unjustifiable. That is, it is impossible to construct a valid serialization consisting of $enq(2)$, $a : deq \triangleright 2$, $enq(1)$, and $b : deq \triangleright 2$ subject to $enq(1) \xrightarrow{vis} a : deq \triangleright 2 \xrightarrow{vis} b : deq \triangleright 2$.

The history of Fig. 4(d) satisfies CM. For example, the return values of $a : p.deq \triangleright 2$ and $b : p.deq \triangleright 1$ can be justified by serializations $\langle q.enq(2) p.enq(1) a : p.deq \triangleright 1 p.enq(2) a : p.deq \triangleright 2 \rangle$ and $\langle p.enq(2) q.enq(2) q.deq \triangleright 2 p.enq(1) b : p.deq \triangleright 2 b : p.deq \triangleright 1 \rangle$, respectively.

5.4 Strong Causal Consistency

We strengthen CM to SCC (Strong Causal Consistency) by further requiring each session to be consistent with respect to the return values provided by other sessions. Formally, we have $V(e) = vis^{-1}(e)$.

Definition 16 (Strong Causal Consistency).

$$SCC \triangleq (hb \subseteq vis) \wedge (vis \subseteq ar) \wedge (V(e) = vis^{-1}(e)) \wedge RVAL.$$

Example 10. The history of Fig. 4(f) satisfies SCC. The return values of $deq \triangleright 1$, $b : deq \triangleright 2$, $b : deq \triangleright 3$, $a : deq \triangleright 3$, and $a : deq \triangleright 2$ can be justified by the serializations $\langle enq(1) deq \triangleright 1 \rangle$, $\langle enq(1) deq \triangleright 1 enq(2) b : deq \triangleright 2 \rangle$, $\langle enq(1) deq \triangleright 1 enq(2) b : deq \triangleright 2 enq(3) deq \triangleright 3 \rangle$, $\langle enq(1) deq \triangleright 1 enq(3) a : deq \triangleright 3 \rangle$ and $\langle enq(1) deq \triangleright 1 enq(3) enq(2) a : deq \triangleright 3 a : deq \triangleright 2 \rangle$, respectively.

The history of Fig. 4(c) does not satisfy SCC. (For now, ignore $enq(2) \xrightarrow{ar} enq(1)$ and $a \xrightarrow{ar} b$ which are for $WCCv$.) Specifically, being aware of the return values of $a : deq \triangleright 2$ and $b : deq \triangleright 2$, $deq \triangleright \perp$ is unjustifiable: it is impossible to construct a valid serialization consisting of all the events, since 2 is dequeued twice.

5.5 Weak Causal Convergence

$WCCv$ (Weak Causal Convergence) [2], [14] is the convergent counterpart of WCC. It strengthens WCC by imposing a total order over all events in an execution, which provides all sessions with a uniform way of resolving conflicts caused by concurrent events. Consequently, the return value $rval(e)$ of each event e is evaluated on the set $vis^{-1}(e)$ of events visible to e , ordered by this common total order ar , while ignoring all of their return values.

Definition 17 (Weak Causal Convergence).

$$WCCv \triangleq (hb \subseteq vis) \wedge (vis \subseteq ar \wedge to(ar, E)) \wedge (V(e) = \emptyset) \wedge RVAL.$$

Example 11. Although the history of Fig. 4(b) satisfies WCC, it does not satisfy $WCCv$. Specifically, the justification for the return value of $b : deq \triangleright 2$ requires

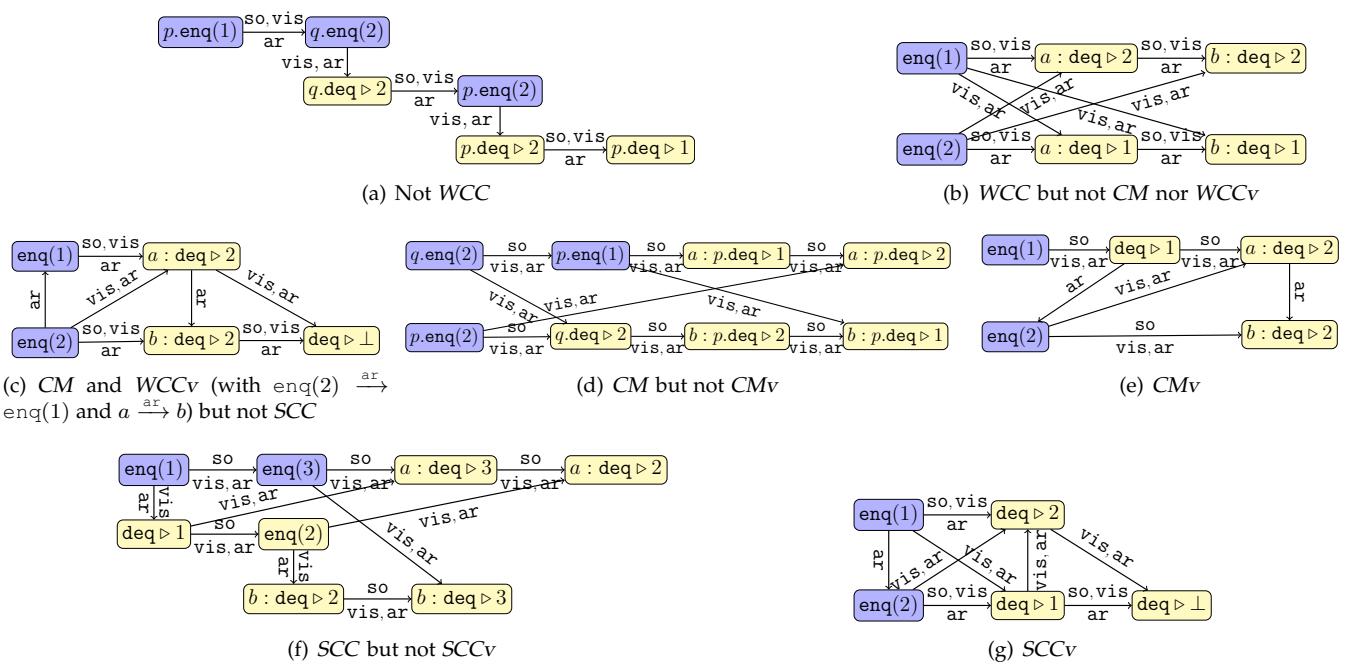


Fig. 4: Examples for causal consistency variants on objects of FIFO queue fq . Both p and q are of type fq in 4(a) and 4(d). The queue q in other subfigures is implicitly assumed. In each history, events are grouped into sessions which are horizontally laid out. The arrows for relations implied by transitivity are not drawn. We use labels, such as a and b , to make events unique.

$\text{enq}(1) \xrightarrow{\text{ar}} \text{enq}(2)$, while the one for $b : \text{deq} \triangleright 1$ requires $\text{enq}(2) \xrightarrow{\text{ar}} \text{enq}(1)$.

The history of Fig. 4(c) satisfies WCCv . The serialization $\langle \text{enq}(2) \text{ enq}(1) a : \text{deq} \triangleright 2 \rangle$ for justifying $a : \text{deq} \triangleright 2$, the one $\langle \text{enq}(2) b : \text{deq} \triangleright 2 \rangle$ for $b : \text{deq} \triangleright 2$, and the one $\langle \text{enq}(2) \text{ enq}(1) a : \text{deq} \triangleright _ b : \text{deq} \triangleright _ \text{deq} \triangleright \perp \rangle$ for $\text{deq} \triangleright \perp$ agree with a common total order ar , e.g., $\langle \text{enq}(2) \text{ enq}(1) a : \text{deq} \triangleright 2 b : \text{deq} \triangleright 2 \text{ deq} \triangleright \perp \rangle$.

5.6 Causal Memory Convergence

CMv (Causal Memory Convergence) is the convergent counterpart of CM , which requires ar to be a total order.

Definition 18 (Causal Memory Convergence).

$$\begin{aligned} \text{CMv} \triangleq \quad & (\text{hb} \subseteq \text{vis}) \wedge (\text{vis} \subseteq \text{ar} \wedge \text{to}(\text{ar}, E)) \\ & \wedge (V(e) = \text{so}^{-1}(e)) \wedge \text{RVAL}. \end{aligned}$$

Example 12. Although the history of Fig. 4(d) satisfies CM , it does not satisfy CMv which enforces a total order ar over all events. As shown in Example 9, the justification for the return value of $a : p.\text{deq} \triangleright 2$ requires $p.\text{enq}(1) \xrightarrow{\text{ar}} p.\text{enq}(2)$, while the justification for the return value of $b : p.\text{deq} \triangleright 1$ requires $p.\text{enq}(2) \xrightarrow{\text{ar}} p.\text{enq}(1)$.

The history of Fig. 4(e) satisfies CMv : the serialization $\langle \text{enq}(1) \text{ deq} \triangleright 1 \rangle$ for justifying the return value of $\text{deq} \triangleright 1$, the one $\langle \text{enq}(1) \text{ deq} \triangleright 1 \text{ enq}(2) a : \text{deq} \triangleright 2 \rangle$ for $a : \text{deq} \triangleright 2$, and the one $\langle \text{enq}(2) b : \text{deq} \triangleright 2 \rangle$ for $b : \text{deq} \triangleright 2$ agree with a common total order ar , e.g., $\langle \text{enq}(1) \text{ deq} \triangleright 1 \text{ enq}(2) a : \text{deq} \triangleright 2 b : \text{deq} \triangleright 2 \rangle$.

5.7 Strong Causal Convergence

SCCv (Strong Causal Convergence) is the convergent counterpart of SCC , which further requires ar to be a total order.

Definition 19 (Strong Causal Convergence).

$$\begin{aligned} \text{SCCv} \triangleq \quad & (\text{hb} \subseteq \text{vis}) \wedge (\text{vis} \subseteq \text{ar} \wedge \text{to}(\text{ar}, E)) \\ & \wedge (V(e) = \text{vis}^{-1}(e)) \wedge \text{RVAL}. \end{aligned}$$

Example 13. Although the history of Fig. 4(f) satisfies SCC , it does not satisfy SCCv . Specifically, the justification for the return value of $a : \text{deq} \triangleright 2$ requires $\text{enq}(3) \xrightarrow{\text{ar}} \text{enq}(2)$, while the one for $b : \text{deq} \triangleright 3$ requires $\text{enq}(2) \xrightarrow{\text{ar}} \text{enq}(3)$.

The history of Fig. 4(g) satisfies SCCv . For $\text{deq} \triangleright \perp$ to return \perp , it must be aware of the event $\text{deq} \triangleright 2$ and by transitivity all other events. It can be justified by the serialization $\langle \text{enq}(1) \text{ enq}(2) \text{ deq} \triangleright 1 \text{ deq} \triangleright 2 \text{ deq} \triangleright \perp \rangle$. So, with $\text{deq} \triangleright 1 \xrightarrow{\text{vis}} \text{deq} \triangleright 2$ and $\text{enq}(1) \xrightarrow{\text{ar}} \text{enq}(2)$, $\text{deq} \triangleright 1$ and $\text{deq} \triangleright 2$ can be justified by the serializations $\langle \text{enq}(1) \text{ enq}(2) \text{ deq} \triangleright 1 \rangle$ and $\langle \text{enq}(1) \text{ enq}(2) \text{ deq} \triangleright 1 \text{ deq} \triangleright 2 \rangle$, respectively.

6 CAUSAL CONSISTENCY: CASE STUDY ON MONGODB

In this section, we prove that the causal consistency protocol of MongoDB [23] satisfies CMv . MongoDB is a distributed database supporting data replication and sharding [23]. It is document-oriented, where a document is an ordered set of key-value pairs. Thus, for simplicity, we model MongoDB as a typical key-value store, which provides $\text{GET}(k)$ and $\text{PUT}(k, v)$ operations to clients. We focus on the causal consistency protocol, called *MongoDB-CC*, of MongoDB in the *failure-free sharded cluster* deployment, where each shard

TABLE 3: Notations used in *MongoDB-CC*.

Notations	Description
ct_c	the greatest cluster time known by client c
ot_c	the last operation time at client c
ct_s	cluster time at server s
aot_s	the last applied operation time at server s
$clock_s$	current physical clock at server s
$oplog_s$	operation log at server s
$\text{CLUSTER}(s)$	the cluster that s belongs to
$\text{PRIMARY}(s)$	the primary node in $\text{CLUSTER}(s)$
$\text{ISPRIMARY}(s)$	whether s is the primary node in $\text{CLUSTER}(s)$
$\text{ISSECONDARY}(s)$	whether s is a secondary node in $\text{CLUSTER}(s)$
k, v	key, value
ct, ot, aot	cluster time, operation time, applied operation time
$oplog$	operation log

is replicated in a cluster consisting of a primary node and several secondary nodes. Only primary nodes can accept PUT operations from clients. Table 3 provides a summary of notations used in the protocol. We reference pseudocode lines using the format algorithm#:line#.

6.1 States

6.1.1 Logical Clocks and Cluster Time

MongoDB-CC uses hybrid logical clocks (HLC) [24]. An HLC is a pair $(sec, counter)$ of physical time (in seconds) and a counter to distinguish operations that occurred within the same second. HLCs are compared lexicographically.

The cluster time is the time value of a node's logical clock [23]. It ticks (Line 4:1) only when a PUT operation is applied in the primary node of a cluster (Line 2:13). Nodes maintain and distribute their greatest known cluster time via messages.

6.1.2 Client and Server States

Each client c keeps track of the greatest known cluster time ct_c . It also maintains the timestamp of its last operation, denoted ot_c , capturing the client's causal dependencies.

Each server s keeps track of the greatest known cluster time ct_s . It uses an append-only operation log $oplog_s$ to record the operations, as well as their timestamps, applied at s . Additionally, it maintains in aot_s the timestamp of the last operation applied at s .

6.2 Protocol

Algorithms 1 and 2 show the core of *MongoDB-CC*, handling GET and PUT operations⁵ at the client and server side, respectively. The pseudocode for replication and clock management are shown in Algorithms 3 and 4, respectively.

6.2.1 GET(k)

A client c sends a GET request, containing the key k , its greatest known cluster time ct_c , and its last operation time ot_c , to a server s which stores key k . The server s first updates its cluster time ct_s (Line 2:2). To guarantee causality, it then checks whether the causal dependencies specified by ot_c have been applied locally, by comparing its last applied

5. The GET and PUT operations are assumed to be executed atomically. The implementation issues about multi-thread concurrency and synchronization are left unspecified.

Algorithm 1 Client operations at client c .

```

1: function GET( $k$ )
2:    $s \leftarrow$  a server storing key  $k$ 
3:    $\langle v, ct, ot \rangle \leftarrow s.\text{GET-REQUEST}(k, ct_c, ot_c)$ 
4:    $ct_c \leftarrow \max(ct_c, ct)$ 
5:    $ot_c \leftarrow ot$                                  $\triangleright t_s(\text{GET}) \leftarrow (ot_c, s), dt(\text{GET}) \leftarrow ot_c$ 
6:   return  $v$ 
7: function PUT( $k, v$ )
8:    $s \leftarrow$  the primary node storing key  $k$ 
9:    $\langle ct, ot \rangle \leftarrow s.\text{PUT-REQUEST}(k, v, ct_c)$ 
10:   $ct_c \leftarrow ct$ 
11:   $ot_c \leftarrow ot$                                  $\triangleright t_s(\text{PUT}) \leftarrow (ot_c, s), dt(\text{PUT}) \leftarrow ot_c$ 
12:  return ok

```

Algorithm 2 Server operations at server s .

```

stores[ $k$ ]  $\leftarrow 0$  for each key  $k$                                  $\triangleright$  Initialization
1: function GET-REQUEST( $k, ct, ot$ )
2:    $ct_s \leftarrow \max(ct_s, ct)$ 
3:   if  $aot_s < ot$  then
4:     PRIMARY( $s$ ).NOOP( $ct_s, ot$ )                                 $\triangleright$  for liveness
5:     if ISSECONDARY( $s$ ) then
6:       repeat
7:          $s.\text{REPLICATE}()$ 
8:       until  $aot_s \geq ot$ 
9:    $v \leftarrow \text{store}_s[k]$ 
10:  return  $\langle v, ct_s, aot_s \rangle$ 
11: function PUT-REQUEST( $k, v, ct$ )
12:    $ct_s \leftarrow \max(ct_s, ct)$ 
13:    $ct_s \leftarrow \text{TICK}()$ 
14:    $aot_s \leftarrow ct_s$ 
15:    $\text{store}_s[k] \leftarrow v$ 
16:    $oplog_s \leftarrow oplog_s \circ \langle k, v, aot_s \rangle$ 
17:   return  $\langle ct_s, aot_s \rangle$                                  $\triangleright ct_s = aot_s$ 

```

operation time aot_s with $ot \leftarrow ot_c$ (Line 2:3). If $aot_s < ot$ and s is a secondary node (Line 2:5), the server keeps replicating $oplog$ from its primary node until $aot_s \geq ot$ (Line 2:8). To ensure liveness, we allow the primary node to catch up by keeping applying NO-OP (Line 4:6) in case $aot_s < ot$ holds in the primary node (Line 2:4). (Note that if $aot_s \geq ot$ and s is the primary node, Line 2:4 does nothing.) Once $aot_s \geq ot$ holds, the server s retrieves the value v of key k in local store_s (Line 2:9). Finally, the value v , as well as ct_s and aot_s are returned to the client (Line 2:10). Upon receiving the reply, the client updates its ct_c and ot_c accordingly.

6.2.2 PUT(k, v)

A client sends a PUT request, containing the key k , the value v , and its greatest known cluster time ct_c , to the server s which is the primary node storing key k . The server s first updates its cluster time ct_s (Line 2:2). Then it ticks its ct_s (Line 2:13) and advances its last applied operation time aot_s to the new ct_s (Line 2:14). After being applied in local store_s (Line 2:15), the PUT operation, as well as its operation time aot_s , is appended to $oplog_s$ (Line 2:16). Finally, both ct_s and aot_s are returned to the client (Line 2:17). Upon receiving the reply, the client updates its ct_c and ot_c accordingly.

6.2.3 Replication

In a cluster, each secondary node s periodically pulls the $oplog$ entries with greater operation time than aot_s from the primary node (Line 3:3). The retrieved entries in $oplog$ are appended to local $oplog_s$ (Line 3:8), and the operations in it are applied in local store_s in increasing order of their

Algorithm 3 Replication at server s .

```

1: function REPLICATE()                                ▷ Run periodically
2:   if ISSECONDARY( $s$ ) then
3:      $\langle oplog, ct \rangle \leftarrow$  PRIMARY( $s$ ).PULL-OPLOG( $ct_s, aot_s$ )
4:      $ct_s \leftarrow \max(ct_s, ct)$ 
5:     for  $\langle k, v, ot \rangle \in oplog$  do                  ▷ in  $ot$  order
6:       store $_s[k] \leftarrow v$ 
7:        $aot_s \leftarrow ot$ 
8:      $oplog_s \leftarrow oplog_s \circ oplog$ 
9: function PULL-OPLOG( $ct, aot$ )
10:    $ct_s \leftarrow \max(ct_s, ct)$ 
11:    $oplog \leftarrow$  oplog entries after  $aot$  in  $oplog_s$ 
12:   return  $\langle oplog, ct_s \rangle$ 

```

Algorithm 4 Clock management at server s .

```

1: function TICK()
2:   if  $ct_s.sec \geq clock_s$  then
3:     return  $\langle ct_s.sec, ct_s.counter + 1 \rangle$ 
4:   else
5:     return  $\langle clock_s, 0 \rangle$ 
6: function NOOP( $ct, ot$ )
7:    $ct_s \leftarrow \max(ct_s, ct)$ 
8:   while  $aot_s < ot$  do
9:      $ct_s \leftarrow$  TICK()
10:     $aot_s \leftarrow ct_s$ 
11:    $oplog_s \leftarrow oplog_s \circ \langle \text{NO-OP}, aot_s \rangle$ 

```

operation times (Line 3:5). In the end, aot_s refers to the operation time of the last entry in its current $oplog_s$ (Line 3:7). Additionally, secondary nodes and the primary node also distribute and keep track of their greatest known cluster time during replication.

6.3 Correctness Proof

We prove that *MongoDB-CC* satisfies *CMv* (Definition 18) by showing that every history $H = (E, \text{op}, \text{rval}, \text{so})$ of *MongoDB-CC* satisfies *CMv* with respect to *kvs* (Example 4). The key is to extract appropriate visibility and arbitration relations from H such that $(H, \text{vis}, \text{ar}) \models \text{CMv}$.

In the following, we use G , P , P_k and G_k to denote the set of GET events, the set of PUT events, the set of PUT events on key k , and the set of GET events on key k , respectively. For a $\text{PUT}(k, v)$ event e , we write $e \triangleright \langle k, v, ot \rangle$ to emphasize that it is associated with an operation time at Line 2:16. For a $\text{GET}(k)$ event e , we write $e \triangleright \langle k, v, ot \rangle$ to denote that it retrieves the value v with operation time ot (Line 2:9).

6.3.1 Timestamps

We first define timestamp $\text{ts}(e)$ for each event e as follows.

Definition 20 (Timestamps). For an event e issued by client c , $\text{ts}(e) = (\text{ot}_c, s)$, where ot_c is the last operation time of client c at Line 1:5 for GET and Line 1:11 for PUT, and s is the identifier of the server processing e (Lines 1:2 and 1:8).

Timestamps are compared lexicographically (we assume a total order over the set of identifiers of servers). For notational convenience, we further define the dependency time $\text{dt}(e) \triangleq \text{ts}(e).\text{ot}_c$ for each event e . Note that for a PUT event e , $\text{dt}(e)$ is its operation time assigned at Line 2:16.

6.3.2 Visibility

The visibility relation vis is based on the following *read-from* relation rf .

Definition 21 (Read-from Relation). $(e, f) \in \text{rf}$ if and only if $e = \text{PUT}(k, v) \triangleright \langle k, v, ot \rangle$ and $f = \text{GET}(k) \triangleright \langle k, v, ot \rangle$ for some key k .

Definition 22 (Visibility). The visibility relation vis is the transitive closure of the union of session order so and read-from relation rf . Formally, $\text{vis} = (\text{so} \cup \text{rf})^+$.

By induction on the structure of vis , we can show that vis is reflected in dt . Formally,

Lemma 1. Let e_1 and e_2 be two events of history H . We have

$$e_1 \xrightarrow{\text{vis}} e_2 \implies \text{dt}(e_1) \leq \text{dt}(e_2).$$

Furthermore,

$$e_1 \xrightarrow{\text{vis}} e_2 \wedge e_2 \in P \implies \text{dt}(e_1) < \text{dt}(e_2).$$

Proof. By induction on the structure of vis .

- CASE I: $e_1 \xrightarrow{\text{so}_c} e_2$ for some client c .
 - ot_c is monotonically increasing. So $\text{dt}(e_1) \leq \text{dt}(e_2)$.
 - If e_2 is a PUT, ot_c is increased due to the ticking at Line 2:13. So $\text{dt}(e_1) < \text{dt}(e_2)$.
- CASE II: $e_1 \xrightarrow{\text{rf}} e_2$.
 - According to the waiting condition at Line 2:3, $\text{ot}_c \leq \text{dt}(e_1)$ always holds when e_2 has been performed at Line 1:5. So $\text{dt}(e_1) \leq \text{dt}(e_2)$.
 - e_2 is a GET.
- CASE III: There is some event e' such that $e_1 \xrightarrow{\text{vis}} e' \xrightarrow{\text{vis}} e_2$.
 - By induction, we have $\text{dt}(e_1) \leq \text{dt}(e') \leq \text{dt}(e_2)$.
 - By induction, if e_2 is a PUT, $\text{dt}(e_1) \leq \text{dt}(e') < \text{dt}(e_2)$.

□

Theorem 2. The visibility relation vis is a partial order.

Proof. It suffices to prove that vis is acyclic (thus, irreflexive). Suppose for a contradiction that there is a cycle $C : e_1 \xrightarrow{\text{vis}} e_2 \xrightarrow{\text{vis}} \dots \xrightarrow{\text{vis}} e_i \xrightarrow{\text{vis}} e_1$. By Lemma 1, all events in C are GET and they have the same dt . Furthermore, all of them cannot occur at the same client, as this would imply a cycle in session order. Assume e and e' in C are on different clients. Since $e \xrightarrow{\text{vis}} e'$ and both of them are GET, there must be some PUT e'' such that $e \xrightarrow{\text{vis}} e'' \xrightarrow{\text{vis}} e'$. By Lemma 1, $\text{dt}(e) < \text{dt}(e'') \leq \text{dt}(e') \leq \text{dt}(e)$, implying $\text{dt}(e) \neq \text{dt}(e')$, a contradiction. □

Theorem 3. $\text{hb} \subseteq \text{vis}$.

Proof. By definitions of hb and vis , we have

$$\text{hb} \triangleq (\text{so} \cup \text{vis})^+ = \text{vis}^+ = \text{vis}.$$

Clearly, $\text{hb} \subseteq \text{vis}$. □

6.3.3 Arbitration

We construct the arbitration relation ar which is a total order over all events as follows.

Definition 23 (Arbitration). Order the PUT events in ar by their timestamps (Definition 20). For each client c , we insert its GET events one by one in session order: each GET event g on client c is placed immediately after the later (in ar) of (1) the previous (in so) event of g on client c , if any; and (2) the PUT event p (in ar) such that $p \xrightarrow{\text{rf}} g$.

Theorem 4. $\text{vis} \subseteq \text{ar}$.

Proof. Let e_1 and e_2 be two events of history H . We need to show that if $e_1 \xrightarrow{\text{vis}} e_2$, then $e_1 \xrightarrow{\text{ar}} e_2$. By induction on the structure of vis .

- CASE I : $e_1 \xrightarrow{\text{so}} e_2$ on the same client. By Condition 1 of ar , $e_1 \xrightarrow{\text{ar}} e_2$.
- CASE II : $e_1 \xrightarrow{\text{rf}} e_2$. By Condition 2 of ar , $e_1 \xrightarrow{\text{ar}} e_2$.
- CASE III : There is some event e' such that $e_1 \xrightarrow{\text{vis}} e' \xrightarrow{\text{vis}} e_2$. By induction and the transitivity of vis and ar , we have $e_1 \xrightarrow{\text{ar}} e' \xrightarrow{\text{ar}} e_2$.

□

6.3.4 Return Values

We need to prove that $(H, \text{vis}, \text{ar}) \models \text{RVAL}(\text{kvs}, V)$, where $V(e) = \text{so}^{-1}(e)$ for each event e . The key is to show, for each GET event e , that

$$\forall e \in G : \text{rval}(e) \in \text{eval}_{\text{kvs}}(\text{ctxt}_A(e, V), \text{op}(e)),$$

where $\text{ctxt}_A(e, V) = A|_{\text{vis}^{-1}(e), \text{op}, \text{rval}|_{\text{so}^{-1}(e)}, \text{vis}, \text{ar}}$. Suppose e is issued by client c . Define $cP \triangleq \text{so}^{-1}(e) \cup (\text{vis}^{-1}(e) \cap P)$ and $S_e \triangleq \text{ar}|_{cP}$.

Theorem 5. S_e is a valid serialization of cP .

Proof. Let $S_{\leq e} \triangleq S_e \circ \langle e \rangle$, where ‘o’ denotes concatenation. We need to show that each GET(k) event in $S_{\leq e}$ returns the value written by the most recently preceding PUT event on key k . Consider $g = \text{GET}(k)$ on client c in $S_{\leq e}$. Suppose $p \xrightarrow{\text{rf}} g$ and p is on client i . We must show that no other PUT($k, _$) events place in between p and g in $S_{\leq e}$. Suppose by contradiction that event $p' = \text{PUT}(k, _)$ on client j does. We distinguish between $j = c$ and $j \neq c$.

- CASE I : $j = c$ (Fig. 5(a)). Since $p \xrightarrow{\text{ar}} p'$, $\text{ts}(p) < \text{ts}(p')$. Since p and p' are applied on the same primary node, $\text{dt}(p) < \text{dt}(p')$. Since $p' \xrightarrow{\text{ar}} g$ and $j = c$, $p' \xrightarrow{\text{so}} g$. So, when g is issued by client c , $\text{ot}_c \geq \text{dt}(p') > \text{dt}(p)$ at Line 1:3. By Line 2:8, it is impossible for g to read from p at Line 2:9.
- CASE II : $j \neq c$. Since $p' \xrightarrow{\text{ar}} g$, there is some event o on client c such that $p' \xrightarrow{\text{ar}} o \xrightarrow{\text{ar}} g$. Let o be the first such event. We perform a case analysis according to whether o is a PUT or a GET.
 - CASE II-1 : $o \in P_k$ (Fig. 5(b)). Since $p \xrightarrow{\text{ar}} p' \xrightarrow{\text{ar}} o$, $\text{ts}(p) < \text{ts}(p') < \text{ts}(o)$. Since both p and p' are applied on the same primary node, $\text{dt}(p) < \text{dt}(p') \leq \text{dt}(o)$. Since $o \xrightarrow{\text{so}} g$, when g is issued by client c , $\text{ot}_c \geq \text{dt}(o) > \text{dt}(p)$ at Line 1:3. By Line 2:8, it is impossible for g to read from p at Line 2:9.
 - CASE II-2 : $o \in G_k$. We consider two cases.

* CASE II-2-A : There are no events between p' and o in $S_{\leq e}$ (Fig. 5(c)). By construction of S_e , $k' = k$ and $p' \xrightarrow{\text{rf}} o$. Therefore, $\text{dt}(p) < \text{dt}(p') \leq \text{dt}(o)$. By a similar argument in CASE II-1, it is impossible for g to read from p .

* CASE II-2-B : There is a set, denoted B , of events between p' and o in $S_{\leq e}$ (Fig. 5(d)). By the choice of o , B contains no events on client c . Therefore, all events in B are PUT. Moreover, there exists some event $p'' \in B$ such that $p'' \xrightarrow{\text{rf}} o$; otherwise, o should be placed before p' in ar . Then, $\text{dt}(p) < \text{dt}(p') \leq \text{dt}(p'') \leq \text{dt}(o)$. By a similar argument in CASE II-1, it is impossible for g to read from p .

□

7 PIPELINED CONSISTENCY: MODELS

7.1 Overview

Pipelined consistency extends the PRAM [12] consistency for shared memory to support arbitrary replicated data types [20]. It requires the updates on a single session are observed by all sessions in the session order they were issued, whereas updates from different sessions may be observed in different orders by different sessions.

In analogous to causal consistency, Table 2 summarizes six variants of pipelined consistency, namely Weak Pipelined Consistency (*WPC*), Pipelined Consistency (*PC*), Strong Pipelined Consistency (*SPC*), Weak Pipelined Convergence (*WPCv*), Pipelined Convergence (*PCv*), and Strong Pipelined Convergence (*SPCv*). Fig. 6 shows the relative strength of these variants and also relates them to their counterparts of causal consistency. These variants of pipelined consistency are weaker than their counterparts of causal consistency, since in the visibility relation they enforce only the session order (i.e., $\text{so} \subseteq \text{vis}$), instead of the happens-before order among events (i.e., $\text{hb} \subseteq \text{vis}$). Fig. 7 gives examples of pipelined consistency on objects of FIFO queues. We explain them in the following subsections.

7.2 Weak Pipelined Consistency

With $V(e) = \emptyset$, *WPC* allows each event e to ignore all the return values of its visible events.

Definition 24 (Weak Pipelined Consistency).

$$\text{WPC} \triangleq (\text{so} \subseteq \text{vis}) \wedge (\text{vis} \subseteq \text{ar}) \wedge (V(e) = \emptyset) \wedge \text{RVAL}.$$

Example 14. The history of Fig. 7(a) does not satisfy *WPC*. Intuitively, events $\text{enq}(1)$, $\text{enq}(3)$, $a : \text{deq} \triangleright 1$, and $\text{deq} \triangleright 3$ must be visible to $b : \text{deq} \triangleright 1$. Since $\text{vis} \subseteq \text{ar}$, it is impossible for $b : \text{deq} \triangleright 1$ to dequeue 1 from the FIFO queue q .

The history of Fig. 7(b) satisfies *WPC*. The return values of $a : \text{deq} \triangleright 1$, $b : \text{deq} \triangleright 1$, and $\text{deq} \triangleright 3$ can be justified by the serializations $\langle \text{enq}(1) \text{ enq}(3) a : \text{deq} \triangleright 1 \rangle$, $\langle \text{enq}(3) \text{ enq}(1) a : \text{deq} \triangleright b : \text{deq} \triangleright 1 \rangle$, and $\langle \text{enq}(1) \text{ enq}(2) \text{ enq}(3) a : \text{deq} \triangleright b : \text{deq} \triangleright \text{deq} \triangleright 3 \rangle$, respectively.

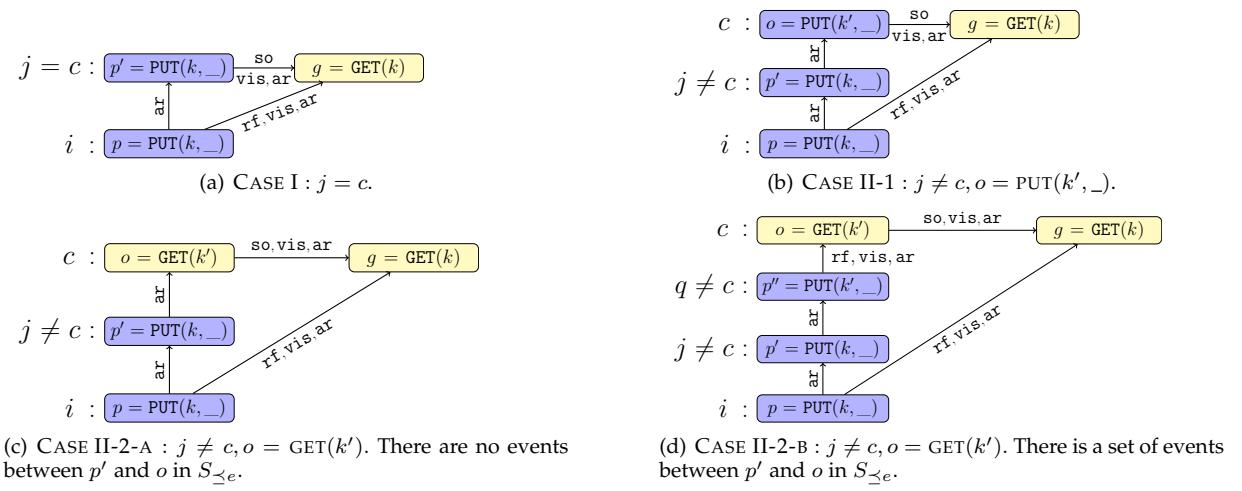


Fig. 5: Illustration of the proof of Theorems 5.

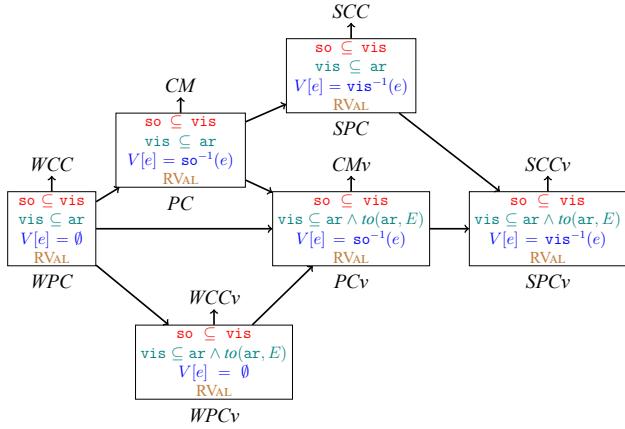


Fig. 6: Pipelined consistency variants and their relative strength.

7.3 Pipelined Consistency

With $V(e) = \text{so}^{-1}(e)$, PC requires that each session is consistent with respect to the previous return values provided.

Definition 25 (Pipelined Consistency).

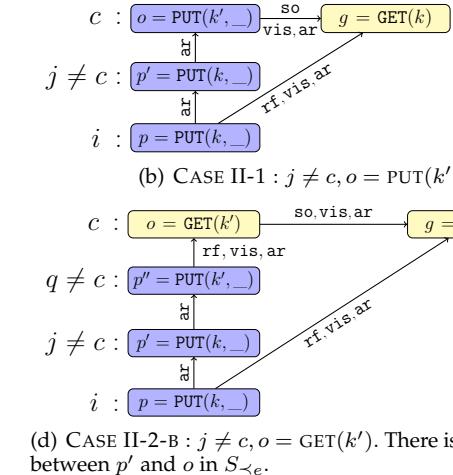
$$PC \triangleq (\text{so} \subseteq \text{vis}) \wedge (\text{vis} \subseteq \text{ar}) \wedge (V(e) = \text{so}^{-1}(e)) \wedge \text{RVAL}$$

Example 15. Although the history of Fig. 7(b) satisfies WPC , it does not satisfy PC . Specifically, being aware of the return value of $a : \text{deq} \triangleright 1, b : \text{deq} \triangleright 1$ is unjustifiable: it is impossible to construct a valid serialization consisting of events $\text{enq}(1), \text{enq}(3), a : \text{deq} \triangleright 1$, and $b : \text{deq} \triangleright 1$, since 1 is dequeued twice.

The history of Fig. 7(d) satisfies PC . For example, the return values of $a : \text{deq} \triangleright 2$ and $b : \text{deq} \triangleright 1$ can be justified by the serializations $\langle \text{enq}(1) \text{ enq}(2) a : \text{deq} \triangleright 1 a : \text{deq} \triangleright 2 \rangle$ and $\langle \text{enq}(2) \text{ enq}(1) b : \text{deq} \triangleright 2 b : \text{deq} \triangleright 1 \rangle$, respectively.

7.4 Strong Pipelined Consistency

With $V(e) = \text{vis}^{-1}(e)$, SPC further requires each session to be also consistent with respect to the return values provided by other sessions.



Definition 26 (Strong Pipelined Consistency).

$$\begin{aligned} SPC \triangleq & (\text{so} \subseteq \text{vis}) \wedge (\text{vis} \subseteq \text{ar}) \\ & \wedge (V(e) = \text{vis}^{-1}(e)) \wedge \text{RVAL}. \end{aligned}$$

Example 16. The history of Fig. 7(c) does not satisfy SPC . (For now, ignore $\text{enq}(1) \xrightarrow{\text{ar}} \text{enq}(2)$ and $b : \text{deq} \triangleright 1 \xrightarrow{\text{ar}} a : \text{deq} \triangleright 1$ which are for $WPCv$.) Specifically, being aware of the return values of $a : \text{deq} \triangleright 1$ and $b : \text{deq} \triangleright 1$, $\text{deq} \triangleright \perp$ is unjustifiable: it is impossible to construct a valid serialization consisting of all the events, since 1 is dequeued twice.

The history of Fig. 7(f) satisfies SPC . The return values of $p.\text{deq} \triangleright \perp$, $q.\text{deq} \triangleright 2$, $q.\text{deq} \triangleright 1$, and $p.\text{deq} \triangleright 3$ can be justified by the serializations $\langle \text{q.enq}(1) p.\text{deq} \triangleright \perp \rangle$, $\langle \text{p.enq}(1) \text{q.enq}(2) q.\text{deq} \triangleright 2 \rangle$, $\langle \text{p.enq}(1) \text{q.enq}(2) q.\text{deq} \triangleright 2 \text{ q.enq}(1) q.\text{deq} \triangleright 1 \rangle$, and $\langle \text{q.enq}(1) p.\text{deq} \triangleright \perp \text{p.enq}(3) \text{p.enq}(1) \text{q.enq}(2) p.\text{deq} \triangleright 3 \rangle$, respectively.

7.5 Weak Pipelined Consistency

With ar being a total order, $WPCv$ is the convergent counterpart of WPC .

Definition 27 (Weak Pipelined Consistency).

$$\begin{aligned} WPCv \triangleq & (\text{so} \subseteq \text{vis}) \wedge (\text{vis} \subseteq \text{ar} \wedge \text{to}(\text{ar}, E)) \\ & \wedge (V(e) = \emptyset) \wedge \text{RVAL}. \end{aligned}$$

Example 17. Although the history of Fig. 7(b) satisfies WPC , it does not satisfy $WPCv$. Specifically, the justification for the return value of $a : \text{deq} \triangleright 1$ requires $\text{enq}(1) \xrightarrow{\text{ar}} \text{enq}(3)$, while the one of $b : \text{deq} \triangleright 1$ requires $\text{enq}(3) \xrightarrow{\text{ar}} \text{enq}(1)$.

The history of Fig. 7(c) satisfies $WPCv$. The serialization $\langle \text{enq}(1) \text{ enq}(2) b : \text{deq} \triangleright 1 \rangle$ for justifying $b : \text{deq} \triangleright 1$, the one $\langle \text{enq}(1) a : \text{deq} \triangleright 1 \rangle$ for justifying $a : \text{deq} \triangleright 1$ and the one $\langle \text{enq}(1) \text{ enq}(2) b : \text{deq} \triangleright \perp a : \text{deq} \triangleright \perp \text{deq} \triangleright \perp \rangle$ for justifying $\text{deq} \triangleright \perp$ agree with a common total order ar , e.g., $\langle \text{enq}(1) \text{ enq}(2) b : \text{deq} \triangleright 1 a : \text{deq} \triangleright 1 \text{ deq} \triangleright \perp \rangle$.

7.6 Pipelined Convergence

With ar being a total order, PCv is the convergent counterpart of PC .

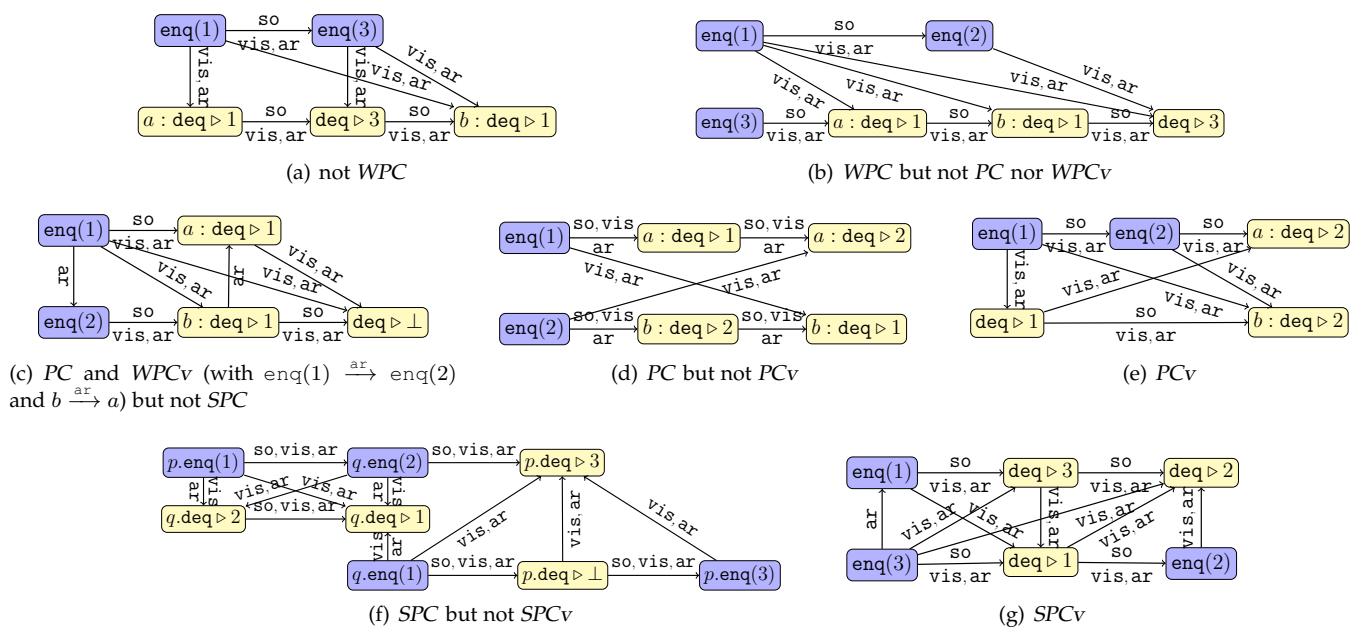


Fig. 7: Examples for pipelined consistency variants on objects of FIFO queue fq. Both p and q are of type fq in 7(f). The queue q in other subfigures is implicitly assumed.

Definition 28 (Pipelined Convergence).

$$\begin{aligned} \text{PCv} \triangleq & \quad (\text{so} \subseteq \text{vis}) \wedge (\text{vis} \subseteq \text{ar} \wedge \text{to}(\text{ar}, E)) \\ & \wedge (V(e) = \text{so}^{-1}(e)) \wedge \text{RVAL}. \end{aligned}$$

Example 18. Although the history of Fig. 7(d) satisfies PC , it does not satisfy PCv which enforces a total order ar over all events. As shown in Example 15, the justification for the return value of $a : \text{deq} \triangleright 2$ requires $\text{enq}(1) \xrightarrow{\text{ar}} \text{enq}(2)$, while the justification for the return value of $b : \text{deq} \triangleright 1$ requires $\text{enq}(2) \xrightarrow{\text{ar}} \text{enq}(1)$.

The history of Fig. 7(e) satisfies PCv . The serialization $\langle \text{enq}(1) \text{ deq} \triangleright 1 \rangle$ for justifying $\text{deq} \triangleright 1$, the one $\langle \text{enq}(1) \text{ enq}(2) \text{ deq} \triangleright a : \text{deq} \triangleright 2 \rangle$ for justifying $a : \text{deq} \triangleright 2$, and the one $\langle \text{enq}(1) \text{ enq}(2) \text{ deq} \triangleright 1 b : \text{deq} \triangleright 2 \rangle$ for justifying $b : \text{deq} \triangleright 2$ agree with a total order, e.g., $\langle \text{enq}(1) \text{ enq}(2) \text{ deq} \triangleright 1 a : \text{deq} \triangleright 2 b : \text{deq} \triangleright 2 \rangle$.

7.7 Strong Pipelined Convergence

With ar being a total order, SPCv is the convergent counterpart of SPC .

Definition 29 (Strong Pipelined Convergence).

$$\begin{aligned} \text{SPCv} \triangleq & \quad (\text{so} \subseteq \text{vis}) \wedge (\text{vis} \subseteq \text{ar} \wedge \text{to}(\text{ar}, E)) \\ & \wedge (V(e) = \text{vis}^{-1}(e)) \wedge \text{RVAL}. \end{aligned}$$

Example 19. Although the history of Fig. 7(f) satisfies SPC , it does not satisfy SPCv . Specifically, the justification for the return value of $q.\text{deq} \triangleright 1$ requires $q.\text{enq}(2) \xrightarrow{\text{ar}} q.\text{enq}(1)$, while the one for $p.\text{deq} \triangleright 3$ requires $q.\text{enq}(1) \xrightarrow{\text{ar}} q.\text{enq}(2)$.

The history of Fig. 7(g) satisfies SPCv . For $\text{deq} \triangleright 2$ to return 2, it must be aware of the event $\text{deq} \triangleright 1$. It can be justified by the serialization $\langle \text{enq}(3) \text{ enq}(1) \text{ deq} \triangleright 3 \text{ deq} \triangleright 1 \text{ enq}(2) \text{ deq} \triangleright 2 \rangle$. So, with $\text{enq}(3) \xrightarrow{\text{ar}} \text{enq}(1)$ and $\text{deq} \triangleright 3 \xrightarrow{\text{vis}} \text{deq} \triangleright 1$, $\text{deq} \triangleright 3$ and $\text{deq} \triangleright 1$ can be justified by serializations $\langle \text{enq}(3) \text{ enq}(1) \text{ deq} \triangleright 3 \rangle$ and $\langle \text{enq}(3) \text{ enq}(1) \text{ deq} \triangleright 3 \text{ deq} \triangleright 1 \rangle$, respectively.

TABLE 4: Notations used in GSP.

Notations	Description
known_c	prefix of total update sequence known by client c
pending_c	local pending updates at client c
round_c	round number at client c
sequence	operation sequence at the server
updates	operations that have not been replicated
k, v, r	key, value, round
c, cl	client
log	operation log

8 PIPELINED CONSISTENCY: CASE STUDY ON GSP

In this section, we show that although the *failure-free client-server* implementation of GSP (Global Sequence Protocol) [25] satisfies WCCv [2, Section 10.4.2], it does not satisfy PCv . GSP requires clients eventually agree on a global sequence of all updates, while seeing a subsequence at any time. To this end, the server uses *Reliable Total Order Broadcast* (RTOB) [26] to reliably deliver messages to all clients in the same total order. We model the system as a key-value store, which supports $\text{GET}(k)$ and $\text{PUT}(k, v)$ operations. Table 4 provides a summary of notations used in the protocol.

8.1 States

The server s maintains a sequence **sequence** of updates received from all clients and a set **updates** of updates that wait to be acknowledged and broadcast.

Each client c maintains two operation sequences, called known_c and pending_c , that stores the currently known prefix of the global update sequence **sequence** and the local pending updates that have not been acknowledged by the server, respectively.

Algorithm 5 Client operations at client c .

```

1: function GET( $k$ )
2:    $log \leftarrow known_c \circ pending_c$ 
3:    $\langle k, v, r, cl \rangle \leftarrow UPDATES(k, log)$ 
4:   return  $v$ 
5: function PUT( $k, v$ )
6:    $round_c \leftarrow round_c + 1$ 
7:    $pending_c \leftarrow pending_c \circ \langle k, v, round_c, c \rangle$ 
8:   send PUT-REQUEST( $k, v, round_c, c$ ) to server  $s$ 
9: function REPLICATE( $k, v, r, cl$ )
10:   $known_c \leftarrow known_c \circ \langle k, v, r, cl \rangle$ 
11:  if  $cl = c$  then
12:     $pending_c \leftarrow pending_c[1..]$ 
```

Algorithm 6 Server operations at server s .

```

1: function PUT-REQUEST( $k, v, r, c$ )
2:    $sequence \leftarrow sequence \circ \langle k, v, r, c \rangle$ 
3:    $updates \leftarrow updates \cup \langle k, v, r, c \rangle$ 
4: function PROPAGATE-UPDATES()            $\triangleright$  Run periodically
5:   for  $\langle k, v, r, c \rangle \in updates$  do
6:     send REPLICATE( $k, v, r, c$ ) to all clients by RTOB
7:    $updates \leftarrow \emptyset$ 
```

8.2 Protocol

To issue a GET operation, a client c combines the updates in $known_c$ and $pending_c$ and determines the return value using an auxiliary function $UPDATES$. To issue a PUT operation, client c increments its local round number $round_c$, appends the update to $pending_c$, and sends it to the server.

When the server s receives a PUT-REQUEST from client c , it appends the PUT operation to $sequence$, and adds it into $updates$. The server periodically replicates the updates in $updates$ to all clients. When a client c receives an update, it appends it to $known_c$. If this update is originated from the client c , it is removed from $pending_c$ since it has been acknowledged by the sever.

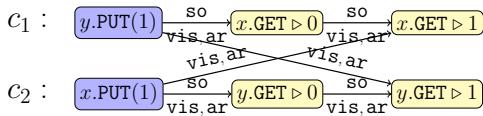


Fig. 8: A history for a kvs that can be produced by the implementation of GSP but does not satisfy PCv.

8.3 The Implementation of GSP does not Satisfy PCv

Consider the history of Figure 8. It consists of two clients c_1 and c_2 which issue GET and PUT operations on keys x and y . The implementation of GSP may produce this history as follows. The sequence at the server is $\langle y.PUT(1) \ x.PUT(1) \rangle$. When $x.GET > 0$ and $x.GET > 1$ are issued on client c_1 , the values of $known_{c_1}$ are $\langle y.PUT(1) \rangle$ and $\langle y.PUT(1) \ x.PUT(1) \rangle$, respectively, and both values of $pending_{c_1}$ are empty. When $y.GET > 0$ and $y.GET > 1$ are issued on client c_2 , the values of $known_{c_2}$ are empty and $\langle y.PUT(1) \rangle$, respectively, and both values of $pending_{c_2}$ are $\langle x.PUT(1) \rangle$.

The history does not satisfies PCv. Specifically, being aware of the return value of $x.GET > 0$, the justification for the return value of $x.GET > 1$ requires $y.PUT(1) \xrightarrow{\text{so,ar}} x.GET > 0 \xrightarrow{\text{ar}} x.PUT(1)$. However, being aware of the return

value of $y.GET > 0$, the justification for the return value of $y.GET > 1$ requires $x.PUT(1) \xrightarrow{\text{so,ar}} y.GET > 0 \xrightarrow{\text{ar}} y.PUT(1)$.

9 PIPELINED CONSISTENCY: CASE STUDY ON RA-LINEARIZABILITY

Replication-Aware linearizability (RA-Linearizability), recently proposed by Wang et al. [27], is a consistency model tailored to Conflict-Free Replicated Data Types (CRDTs [28], [29]). RA-Linearizability relaxes linearizability [30] in two ways: (1) the linearization is required to be consistent with visibility relation instead of the real-time *returns-before* relation; (2) query events are allowed to see a subsequence, rather than a prefix, of the linearization. In this section, we show that RA-Linearizability implies WPCv, but not PCv.

Definition 30 (RA-Linearizability). Let $h = (L, vis)$ be a history of a CRDT⁶, where L is a set of Queries and Updates events⁷ and vis is a partial order over L . The history h satisfies RA-Linearizability w.r.t. a sequential specification $Spec$, if there exists a sequence seq such that

- seq is a total order over L ;
- $so \subseteq vis$ (we use sessions to denote the replicas);
- $vis \cup seq$ is acyclic; and
- for each query l_q in L , the subsequence of updates visible to l_q together with l_q , i.e., $seq|_{vis^{-1}(l_q)} \cap Updates$, is admitted by $Spec$.

9.1 RA-Linearizability Implies WPCv

The definition of RA-Linearizability is “isomorphic” to that of WPCv, in terms of the following three definitions.

Definition 31 (Visibility). $vis \triangleq vis$.

Definition 32 (Arbitration). $ar \triangleq seq$.

Definition 33. $\text{eval}_{\text{CRDT}} \triangleq Spec$.

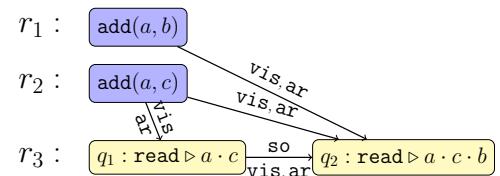


Fig. 9: A history for a list object l (implicitly assumed) that satisfies RA-Linearizability but not PCv.

9.2 RA-Linearizability does not Imply PCv

Consider the history of Figure 9. It consists of three replicas r_1 , r_2 , and r_3 which perform operations on a list l of characters. The list supports two operations⁸: (1) $\text{add}(a, b)$ which inserts the character b immediately after the character a in the list, and (2) read which returns the list content. Assume that the list l initially contains only the character a .

6. RA-Linearizability is compositional under some assumptions; see [27, Section 5].

7. We do not consider query-update events and query-update rewriting; see [27, Definition 3.6].

8. We do not consider the *remove* operation in this example.

The history satisfies RA-Linearizability. The total sequence seq can be $\langle \text{add}(a, b) \text{ add}(a, c) q_1 : \text{read} \triangleright a \cdot c q_2 : \text{read} \triangleright a \cdot c \cdot b \rangle$. With $\text{add}(a, c) \xrightarrow{\text{vis}} q_1$, $\text{add}(a, b) \xrightarrow{\text{vis}} q_2$, and $\text{add}(a, c) \xrightarrow{\text{vis}} q_2$, the return values of q_1 and q_2 can be justified by the subsequences $\langle \text{add}(a, c) \rangle$ and $\langle \text{add}(a, b) \text{ add}(a, c) \rangle$, respectively.

However, the history does *not* satisfy PCv . Being aware of the return value of q_1 , q_2 is unjustifiable: q_1 requires $\text{add}(a, b)$ be ordered after both $\text{add}(a, c)$ and q_1 , but q_2 requires $\text{add}(a, b)$ be ordered before $\text{add}(a, c)$.

10 RELATED WORK

The (vis, ar) specification framework for arbitrary eventually consistent replicated data types is proposed by Burckhardt et al. [1], [2]. It introduces the visibility and arbitrary relations, which have been widely adopted in the literature. Using this framework, Viotti et al. [10] provide a comprehensive overview of more than 50 different consistency models in distributed systems. Emmi et al. [31] develop a fine-grained consistency specification methodology for software API via visibility relaxation. Recently, Almeida et al. [16] proposed a framework for consistency models in distributed systems using the traditional per-process serializations augmented with visibility. In this paper, we extend the (vis, ar) framework into a more generic one called (vis, ar, V) for weakly consistent replicated data types. Our framework is able to cover not only existing consistency models but also new ones that are reasonable and promising for practical usefulness.

Several works develop uniform frameworks for consistency models in *shared-memory* multiprocessor systems. Steinke et. al [19] present a unified theory of shared-memory consistency models based on four consistency properties. Enumerating all combinations of these four properties produces a lattice of consistency models. Alglave [32] provides a generic framework for weak consistency models in modern multiprocessor architectures. It uses the global time model and addresses the store atomicity relaxation.

Sequential consistency [21] is considered a strong consistency model. It is used in coordination services, such as ZooKeeper [33]. Interestingly, in our (vis, ar, V) framework, three variants of sequential consistency with different V functions turn out to be equivalent.

Causal consistency has been widely used in distributed systems [13], [22], [34]. There are three known causal consistency variants in the literature, namely WCC [14], [15], CM [13], [14], [15], and $WCCv$ [2], [14], [15]. By following the recipes for our framework, we discover three new causal consistency variants, namely SCC , CMv , and $SCCv$. We show that the causal consistency protocol of MongoDB [23] satisfies CMv . As far as we know, this is the first correctness proof for MongoDB protocols against formal specifications.

Pipelined consistency [20] extends the PRAM consistency [12] in shared-memory multiprocessor systems to support arbitrary replicated data types in distributed systems. $WPCv$ is a convergent variant of pipelined consistency [20]. We discover four new pipelined consistency variants, namely WPC , SPC , PCv , and $SPCv$. We show that the implementation of GSP [25] does not satisfy PCv , and

that the RA-Linearizability [27] specification implies $WPCv$ but not PCv .

11 CONCLUSION AND FUTURE WORK

We extend the (vis, ar) specification framework for eventually consistent replicated data types into a more generic one called (vis, ar, V) for weakly consistent replicated data types. It covers both non-convergent consistency models and the consistency models in which each event is required to be aware of the return values of some or all events that are visible to it. We also provide case studies on MongoDB, the GSP protocol, and the RA-Linearizability specification to demonstrate the usefulness of our framework. We are implementing our specification framework in Alloy [35].

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