Reconfigurable Lattice Agreement and Applications

Petr Kuznetsov, Thibault Rieutord, Sara Tucci-Piergiovanni

Reconfigurable Lattice Agreement and Applications

Petr Kuznetsov
LTCI, Télécom Paris, Institut Polytechnique Paris
petr.kuznetsov@telecom-paris.fr

Thibault Rieutord
CEA LIST, PC 174, Gif-sur-Yvette, 91191, France
thibault.rieutord@cea.fr

Sara Tucci-Piergiovanni
CEA LIST, PC 174, Gif-sur-Yvette, 91191, France
sara.tucci@cea.fr

Abstract

Reconfiguration is one of the central mechanisms in distributed systems. Due to failures and connectivity disruptions, the very set of service replicas (or *servers*) and their roles in the computation may have to be reconfigured over time. To provide the desired level of consistency and availability to applications running on top of these servers, the *clients* of the service should be able to reach some form of agreement on the system configuration. We observe that this agreement is naturally captured via a *lattice* partial order on the system states. We propose an asynchronous implementation of *reconfigurable* lattice agreement that implies elegant reconfigurable versions of a large class of *lattice* abstract data types, such as max-registers and conflict detectors, as well as popular distributed programming abstractions, such as atomic snapshot and commit-adopt.

2012 ACM Subject Classification Theory of computation → Design and analysis of algorithms → Distributed algorithms

Keywords and phrases Reconfigurable services, lattice agreement

Digital Object Identifier 10.4230/LIPIcs...

1 Introduction

A decentralized service [6, 14, 24, 27] runs on a set of fault-prone *servers* that store replicas of the system state and run a synchronization protocol to ensure consistency of concurrent data accesses. In the context of a storage system exporting read and write operations, several proposals [2, 3, 18, 20, 23, 30] came out with a reconfiguration interface that allows the servers to join and leave, while ensuring consistency of the stored data. Early proposals [20] were based on using *consensus* [16, 21] to ensure that replicas *agree* on the evolution of the system membership. Consensus, however, is expensive and difficult to implement, and recent solutions [2, 3, 18, 23, 30] replace consensus with weaker abstractions capturing the minimal coordination required to safely change the servers configuration. These solutions, however, lack of a uniform way of deriving reconfigurable versions of static objects.

Reconfiguration lattices. In this paper, we propose a universal construction for a large class of objects. Unlike a consensus-based reconfiguration proposed earlier for generic state-machine replication [25], our construction is asynchronous, at the expense of assuming a restricted object behavior. More precisely, we assume that the set $\mathcal{L}$ of the object's states can be represented as a (join semi-) *lattice* $(\mathcal{L}, \sqsubseteq)$, where $\mathcal{L}$ is partially ordered by the binary relation $\sqsubseteq$ such that for all elements of $x, y \in \mathcal{L}$, there exists the *least upper bound* in $\mathcal{L}$,

denoted $x \sqcup y$, where $\sqcup$ is an associative, commutative, and idempotent binary operator on $\mathcal{L}$. Many important data types, such as atomic snapshots, sets and counters, as well as useful concurrent abstractions, such as commit-adopt [17], can be expressed this way. Intuitively, $x \sqcup y$ can be seen as a *merge* of two alternatively proposed updated states x and y. As long as an implementation of the object ensures that all "observable" states are ordered by $\sqsubseteq$, it cannot be distinguished from an atomic object.

Consider, for example, the *max-register* [4] data type which exports two operations: *writeMax* that writes values and *readMax* that returns the largest value written so far. Its state space can be represented as a lattice $(\sqsubseteq, \sqcup)$ of its values, where $\sqsubseteq = \leqslant$ and $x \sqcup y = \max(x, y)$. Intuitively, a linearizable concurrent implementation of max-register must ensure that every read value is a join of previously proposed values, and all read values are totally ordered (with respect to $\sqsubseteq$).

Reconfigurable lattice agreement. The observation above inspires an elegant approach to build reconfigurable objects. In this paper, we introduce reconfigurable lattice agreement [8, 15]. It is natural to treat the *system configuration*, i.e., the set of servers available for data replication, as an element in a lattice. A lattice-defined merge of configurations, possibly concurrently proposed by different processes, results in a new configuration. The lattice-agreement protocol ensures that configurations evaluated by concurrent processes are *ordered*. Despite processes possibly disagreeing about the precise configuration they belong to, they can use these diverging configurations to safely implement lattice agreement.

We assume that a configuration is a set of servers provided with a quorum system [19], i.e., a set system ensuring the intersection property[1] and, possibly, other configuration parameters. For example, elements of a reconfiguration lattice can be defined as sets of *configuration updates*: each such update either adds a server to the configuration or removes a server from it. The *members* of a configuration are the set of all servers that were added but not yet removed. A join of two configurations defined this way is simply a union of their updates (this approach is implicitly used in earlier asynchronous reconfigurable constructions [2, 18, 30]).

Reconfigurable L-ADT and applications. We show that our reconfigurable lattice agreement, defined on a product of a *configuration lattice* and an *object lattice*, immediately implies reconfigurable versions of many sequential types, such as *max-register* and *conflict detector*. More generally, any *state-based commutative* abstract data (called *L-ADT*, for *lattice abstract data type*, in this paper) has a reconfigurable *interval-linearizable* [12] implementation. Intuitively, interval-linearizability [12], a generalization of the classical linearizability [22], allows to specify the behavior of an object when multiple concurrent operations "influence" each other. Their effects are then merged using a join operator, which turns out to be natural in the context of reconfigurable objects.

Our transformations are straightforward. To get an (interval-linearizable) reconfigurable implementation of an L-ADT, we simply use its state lattice, as a parameter, in our reconfigurable lattice agreement. The resulting implementations are naturally composable: we get a reconfigurable composition of two L-ADTs by using a *product* of the their lattices. When operations on the object can be partitioned into *updates* (modifying the object state without providing informative responses) and *queries* (not modifying the object state), as in the case of max-registers, the reconfigurable implementation is also linearizable[2].

[1] The most commonly used quorum system is majority-based: quorums are all majorities of servers. We can, however, use any other quorum system, as suggested in [20, 23].

[2] This class of "update-query" L-ADTs is known as state-based convergent replicated data types (CvRDT) [28]. These types include *max-register*, *set* and *abort flag* (a new type introduced in this

We then use our reconfigurable implementations of *max-register*, *conflict detector*, *set* and *abort-flag* to devise reconfigurable versions of *atomic snapshot* [1], *commit-adopt* [17] and *safe agreement* [10]. Figure 1 shows how are constructions are related.

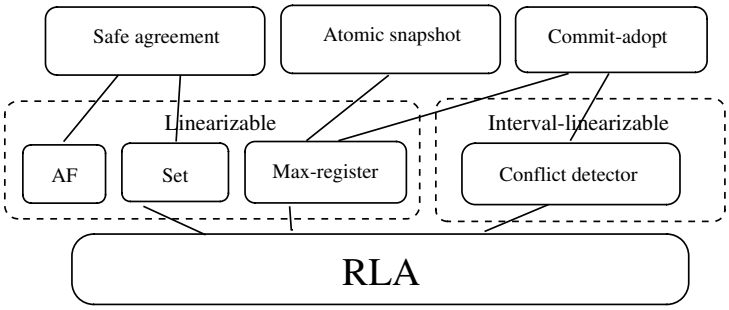

Figure 1 Our reconfigurable implementations: reconfigurable lattice agreement (RLA) is used to construct linearizable implementations of a set, a max-register, an abort flag, and an interval-linearizable implementation of a conflict detector. On top of max-registers we construct an atomic snapshot, on top of max-registers and conflict detector, we construct a commit-adopt abstraction, and on top of set and abort flag, we implement safe agreement.

Summary. Our reconfigurable construction is the first to be, at the same time:

- Asynchronous, unlike consensus-based solutions [13,20,25], and not assuming an external lattice agreement service [23];
- Uniformly applicable to a large class of objects, unlike existing reconfigurable systems that either focus on read-write storage [2, 18, 20, 23] or require data type-specific implementations of exported reconfiguration interfaces [30];
- Allowing for a straightforward composition of reconfigurable objects;
- Maintaining configurations with abstract *quorum systems* [19], not restricted to *majority-based* quorums [2, 18];
- Exhibiting optimal time complexity and message complexity comparable with the best known implementations [2, 23, 30];
- Logically separating *clients* (external entities that use the implemented service) from *servers* (entities that maintain the service and can be reconfigured).

We also believe our reconfigurable construction to be the simplest on the market, using only twenty two lines of pseudocode and provided with a concise proof.

Roadmap. The rest of the paper is organized as follows. We give basic model definitions in Section 2. In Section 3, we define our type of reconfigurable objects, followed by the related notion of reconfigurable lattice agreement in Section 4. In Section 5, we describe our implementation of reconfigurable lattice agreement, and, in Section 6, we show how to use it to implement a reconfigurable L-ADT object. In Section 7 we describe some possible applications. Then, we conclude in Section 8 with an overview of the related work.

2 Definitions

Replicas and clients. Let Π be a (possibly infinite) set of potentially participating processes. A subset of the processes, called *replicas*, are used to maintain the implemented

paper).

object. A process can also act as a *client*, proposing operations on the implemented object and system reconfigurations. Replicas and clients are subject to crash faults: a process *fails* when it prematurely stops taking steps of its algorithm. A process is *correct* if it never fails.

Abstract data types. An abstract data type (ADT) is defined as a tuple $T = (A, B, Z, z_0, \tau, \delta)$. Here A and B are countable sets called the *inputs* and *outputs*. Z is a countable set of abstract object *states*, $z_0 \in Z$ being the initial state of the object. The map $\tau : Z \times A \to Z$ is the *transition function*, specifying the effect of an input on the object state and the map $\delta : Z \times A \to B$ is the *output function*, specifying the output returned for a given input and object local state. The input represents an operation with its parameters, where (i) the operation can have a side-effect that changes the abstract state according to transition function τ and (ii) the operation can return values taken in the output B, which depends on the state in which it is called and the output function δ (for simplicity, we only consider deterministic types here, check, e.g., [26], for more details.)

Interval linearizability. We now briefly recall the notion of *interval-linearizability* [12], a recent generalization of linearizability [22].

Let us consider an abstract data type $T = (A, B, Z, z_0, \tau, \delta)$. A *history* of T is a sequence of inputs (elements of A) and outputs (elements of B), each labeled with a process identifier and an operation identifier. An *interval-sequential history* is a sequence:

$$z_0, I_1, R_1, z_1, I_2, R_2, z_2 \ldots, I_m, R_m, z_m,$$

where each $z_i \in Z$ is a state, $I_i \subseteq A$ is a set of inputs, and $R_i \subseteq B$ is a set of outputs. An *interval-sequential specification* is a set of interval-sequential histories.

We only consider *well-formed* histories. Informally, in a well-formed history, a process only invokes an operation once its previous operation has returned and every response r is preceded by a "matching" operation i.

A history H is *interval-linearizable* respectively to an interval-sequential specification $\mathcal{S}$ if it can be *completed* (by adding matching responses to incomplete operations) so that the resulting history $\bar{H}$ can be associated with an interval-sequential history S such that: (1) $\bar{H}$ and S are *equivalent*, i.e., $\forall p \in \Pi$, $\bar{H}|p = S|p$, (2) $S \in \mathcal{S}$, and (3) $\to_H \subseteq \to_S$, i.e., S preserves the real-time precedence relation of H. (Check [12] for more details on the definition.)

Lattice agreement. An abstract (join semi-)*lattice* is a tuple $(\mathcal{L}, \sqsubseteq)$, where $\mathcal{L}$ is a set partially ordered by the binary relation $\sqsubseteq$ such that for all elements of $x, y \in \mathcal{L}$, there exists the least upper bound for the set $\{x, y\}$. The least upper bound is an associative, commutative, and idempotent binary operation on $\mathcal{L}$, denoted by $\sqcup$ and called the *join operator* on $\mathcal{L}$. We write $x \sqsubset y$ whenever $x \sqsubseteq y$ and $x \neq y$. With a slight abuse of notation, for a set $L \subseteq \mathcal{L}$, we also write $\bigsqcup L$ for $\bigsqcup_{x \in L} x$, i.e., $\bigsqcup L$ is the join of the elements of L.

Notice that two lattices $(\mathcal{L}_1, \sqsubseteq_1)$ and $(\mathcal{L}_2, \sqsubseteq_2)$ naturally imply a *product* lattice $(\mathcal{L}_1 \times \mathcal{L}_2, \sqsubseteq_1 \times \sqsubseteq_2)$ with a product join operator $\sqcup = \sqcup_1 \times \sqcup_2$. Here for all $(x_1, x_2), (y_1, y_2) \in \mathcal{L}_1 \times \mathcal{L}_2$, $(x_1, x_2)(\sqsubseteq_1 \times \sqsubseteq_2)(y_1, y_2)$ if and only if $x_1 \sqsubseteq_1 y_1$ and $x_2 \sqsubseteq_2 y_2$.

The (generalized) *lattice agreement* concurrent abstraction, defined on a lattice $(\mathcal{L}, \sqsubseteq)$, exports a single operation *propose* that takes an element of $\mathcal{L}$ as an argument and returns an element of $\mathcal{L}$ as a response. When the operation *propose*(x) is invoked by process p we say that p *proposes* v, and when the operation returns v' we say that p *learns* v'. Assuming that no process invokes a new operation before its previous operation returns, the abstraction satisfies the following properties:

- **Validity.** If a *propose*(v) operation returns a value v' then v' is a join of some proposed values including v and all values learnt before the invocation of the operation.

- **Consistency.** The learnt values are totally ordered by $\sqsubseteq$.
- **Liveness.** Every *propose* operation invoked by a correct process eventually returns.

A historical remark. The original definition of long-lived lattice agreement [15] separates "receive" events and "learn" events. Here we suggest a simpler definition that represents the two events as the invocation and the response of a *propose* operation. This also allows us to slightly strengthen the validity condition so that it accounts for the *precedence* relation between *propose* operations. As a result, we can directly relate lattice agreement to linearizable [22] and interval-linearizable [12] implementations, without introducing artificial "nop" operations [15].

3 Lattice Abstract Data Type

In this section, we introduce a class of types that we call *lattice abstract data types* or *L-ADT*. In an *L-ADT*, the set of states forms a join semi-lattice with a partial order $\sqsubseteq^Z$. A lattice object is therefore defined as a tuple $L = (A, B, (Z, \sqsubseteq^Z, \sqcup^Z), z_0, \tau, \delta)$.[3] Moreover, the transition function δ must comply with the partial order $\sqsubseteq^Z$, that is $\forall z, a \in Z \times A : z \sqsubseteq^Z \tau(z, a)$, and the composition of transitions must comply with the join operator, that is $\forall z \in Z, \forall a, a' \in A : \tau(\tau(z, a), a') = \tau(z, a) \sqcup^Z \tau(z, a') = \tau(\tau(z, a'), a)$. Hence, we can say that the transition function is "commutative".

Update-query L-ADT. We say an L-ADT $L = (A, B, (Z, \sqsubseteq^Z, \sqcup^Z), z_0, \tau, \delta)$ is *update-query* if A can be partitioned in *updates* U and *queries* Q such that:

- there exists a special "dummy" response $\bot$ (z_0 may also be used) such that $\forall u \in U,\ z \in Z,\ \delta(u, z) = \bot$, i.e., updates do not return informative responses;
- $\forall q \in Q,\ z \in Z,\ \tau(u, z) = z$, i.e., queries do not modify the states.

This class of types is also known as a state-based convergent replicated data types (CvRDT) [28]. Typical examples of update-query L-ADTs are *max-register* [4] (see Section 1) or *sets*. Note that any (L-)ADT can be transformed into an update-query (L-)ADT by "splitting its operations" into an update and a query (see [26]).

Composition of L-ADTs. The composition of two ADTs $T = (A, B, Z, z_0, \tau, \delta)$ and $T' = (A', B', Z', z'_0, \tau', \delta')$ is denoted $T \times T'$ and is equal to $(A + A', B \cup B', Z \times Z', (z_0, z'_0), \tau'', \delta'')$; where $A + A'$ denotes the disjoint union and where τ'' and δ'' apply, according to the domain A or A' of the input, either τ and δ or τ' and δ' on their respecting half of the state (see [26]).

Since the cartesian product of two lattices remains a lattice, the composition of L-ADTs is naturally defined and produces an L-ADT. The composition is also closed to update-query ADT, and thus to update-query L-ADT. Moreover, the composition is an associative and commutative operator, and hence, can easily be used to construct elaborate L-ADT.

Configurations as L-ADTs. The reconfiguration service can similarly defined as follows. Let us define a *configuration L-ADT* as a tuple $(A^{\mathcal{C}}, B^{\mathcal{C}}, (\mathcal{C}, \sqsubseteq^{\mathcal{C}}, \sqcup^{\mathcal{C}}), C_0, \tau^{\mathcal{C}}, \delta^{\mathcal{C}})$. For each element C of the *configuration lattice* $\mathcal{C}$, the input set A includes the operation *members*(), such that $\delta^{\mathcal{C}}(C, members()) \subseteq \Pi$, and the operation *quorums*() such that $\delta^{\mathcal{C}}(C, quorums())$ is $\subseteq 2^{\delta^{\mathcal{C}}(C, members())}$, a *quorum system*, where every two subsets in $\delta^{\mathcal{C}}(C, quorums())$ have a non-empty intersection. In the following we will denote these two operations, with a slight abuse of notation, as $members(C)$ and $quorums(C)$. Here C_0 is called the *initial configuration*.

[3] For convenience, we explicitly specify the join operator $\sqcup^Z$ here, i.e., the least upper bound of $\sqsubseteq^Z$.

For example, $\mathcal{C}$ can be the set of tuples (In, Out), where $In \subseteq \Pi$ is a set *activated* processes, and $Out \subseteq \Pi$ is a set of *removed* processes. Then $\sqsubseteq^{\mathcal{C}}$ can be defined as the piecewise set inclusion on (In, Out). The set of members of (In, Out) will simply be $In - Out$ and the set of quorums (pairwise-intersecting subsets of $In - Out$), e.g., all majorities of $In - Out$. Operations in $A^{\mathcal{C}}$ can be $add(s)$, $s \in \Pi$, that adds s to the set of activated processes and $remove(s)$, $s \in \Pi$, that adds s to the set of removed processes of a configuration. One can easily see that updates "commute" and that the type is indeed an L-ADT. Let us note that L-ADTs allow for more expressive reconfiguration operations than simple *adds* and *removes*, e.g., maintaining a minimal number of members in a configuration or adapting the quorum system dynamically, as studied in detail by Jehl et al. in [23].

Interval-sequential specifications of L-ADTs. Let $L = (A, B, (Z, \sqsubseteq^Z, \sqcup^Z), z_0, \tau, \delta)$ be an L-ADT. As τ "commutes", the state reached after a sequence of transitions is order-independent. Hence, we can define a natural interval-sequential specification of L, $\mathcal{S}_L$, as the set of interval-sequential histories $z_0, I_1, R_1, z_1, I_2, R_2, z_2, \ldots, I_m, R_m, z_m$ such that:

- $\forall i = 1, \ldots, m$, $z_i = \bigsqcup^Z_{a \in I_{i-1}} \tau(a, z_{i-1})$, i.e., every state z_i is a join of operations in I_{i-1} applied to z_{i-1}.
- $\forall i = 1, \ldots, m$, $\forall r \in R_i$, $r = \delta(a, z_i)$, where a is the matching invocation operation for r, i.e., every response in R_i is based on the result of the corresponding input applied to state z_i.

4 Reconfigurable lattice agreement: definition

We define a reconfigurable object as a composition of two L-ADTs, an *object* L-ADT $(A^{\mathcal{O}}, B^{\mathcal{O}}, (\mathcal{O}, \sqsubseteq^{\mathcal{O}}, \sqcup^{\mathcal{O}}), O_0, \tau^{\mathcal{O}}, \delta^{\mathcal{O}})$ and a *configuration* L-ADT $(A^{\mathcal{C}}, B^{\mathcal{C}}, (\mathcal{C}, \sqsubseteq^{\mathcal{C}}, \sqcup^{\mathcal{C}}), C_0, \tau^{\mathcal{C}}, \delta^{\mathcal{C}})$ (see Section 3). Our main tool is the reconfigurable lattice agreement, a generalization of lattice agreement operating on the product $(\mathcal{L}, \sqsubseteq) = (\mathcal{O} \times \mathcal{C}, \sqsubseteq^{\mathcal{O}} \times \sqsubset^{\mathcal{C}})$ with the product join operator $\sqcup = \sqcup^{\mathcal{O}} \times \sqcup^{\mathcal{C}}$. We say that $\mathcal{L}$ is the set of *states*. For a state $u = (O, C) \in \mathcal{L}$, we use notations $u.O = O$ and $u.C = C$.

When a process p invokes $propose((O, C))$, we say p *proposes* object state O and configuration $\{C\} \in \mathcal{C}$.

We say that p *learns* an object state O' and a configuration C' if its *propose* invocation returns (O', C').

The idea is to maintain replicas of a reconfigurable object on active members of installed but not yet superseded configurations. Formally, we say that a proposed configuration C is *installed* as soon as some process learns $(*, C')$ such that $C \sqsubseteq^{\mathcal{C}} C'$. A configuration C is *available* if some set in $quorums(C)$ contains only correct processes. A configuration is *superseded* as soon some process learns a state $(*, C')$ such that $C \sqsubseteq^{\mathcal{C}} C'$ and $C \neq C'$.

In a constantly reconfigured system, we may not be able to ensure liveness to all operations. A slow client can be always behind the installed and not superseded configuration: the set of servers it believes to be currently active can always be found to constitute a superseded configuration. Therefore, for liveness, we assume that only finitely many reconfigurations occur.

Moreover, we require that any join of proposed configurations that is never superseded must be available:

- **Configuration availability.** Let $C_1, \ldots, C_k$ be proposed configurations such that $C = \bigsqcup^{\mathcal{C}}_{i=1,\ldots,k} C_i$ is never superseded. Then C is available.

Therefore, any configuration constructed as a join of proposed configurations and"superseded" by a strictly larger (w.r.t. $\sqsubseteq^{\mathcal{C}}$) configuration does not have to be available, so it can safely remove some servers for maintenance. In the rest of the paper, we implicitly assume configuration availability in arguing liveness.

As a client may not be aware of the current installed and not superseded configuration, we can only guarantee liveness to slow clients assuming that, eventually, every *correct* system participant (client or replica) is informed of the currently active configuration. Here we need to amend the notion of a correct process, having a reconfigurable system in mind.

We say a replica joins the system when the first configuration it belongs to is proposed, and leaves the system when the first configuration it does not belong to is learnt. Now a replica is called *correct* if it joined the system and never failed or left. A *client* is correct if it does not fail while executing its propose operation.

We assume that a reliable broadcast primitive [11] is available, ensuring that (i) if a correct process broadcasts a message, then it eventually delivers it and (ii) every message delivered by a correct process is eventually delivered by every correct process.

To get a reconfigurable object, we therefore replace the liveness property of lattice agreement with the following one:

- **Reconfigurable Liveness.** In executions with finitely many distinct proposed configurations, every *propose* operation invoked by a correct client eventually returns.

Note that the desired liveness guarantees are ensured as long as only finitely many distinct *configurations* are proposed. However, the clients are free to perform infinitely many *object* updates without making any correct process starve.

Formally, *reconfigurable lattice agreement* defined on $(\mathcal{L}, \sqsubseteq) = (\mathcal{O} \times \mathcal{C}, \sqsubseteq^{\mathcal{O}} \times \sqsubset^{\mathcal{C}})$ satisfies the Validity and Consistency properties of lattice agreement (see Section 2) and the Reconfigurable Liveness property above.

5 Reconfigurable lattice agreement: implementation

We now present our main technical result, a reconfigurable implementation of generalized lattice agreement. This algorithm will then be used to implement reconfigurable objects.

Overview. The algorithm is specified by the pseudocode of Figure 2. Note that we assume that all procedures (including sub-calls to the *updateState* procedure) are executed *sequentially* until they terminate or get interrupted by the wait condition in line 9.

In the algorithm, every process (client or server) p maintains a *state* variable $v_p \in \mathcal{L}$ storing its local estimate of the greatest committed object ($v_p.O$) and configuration ($v_p.C$) states, initialized to the initial element of the lattice (O_0, C_0). We say that a state is committed if a process broadcasted it in line 13. Note that all learnt states are committed (possibly indirectly by another process), but a process may fail before learning its committed value. Every process p also maintains T_p, the set of *active input* configuration states, i.e., input configuration states that are not superseded by the committed state estimate v_p. For the object lattice, processes stores in obj_p the join of all known proposed objects states.

To propose *prop*, client p update its local variables using the *updateState* procedure using its input object and configuration states, *prop.O* and *prop.C* (line 1). Clients then enter a while loop where they send *requests* associated with their current sequence number seq_p and containing the triplet (v_p, obj_p, T_p), to all replicas from *every possible join* of active base configurations and wait until either (1) they get interrupted by discovering a greater committed configuration through the underlying reliable broadcast, or (2) for each possible

join of active base configurations, a quorum of its replicas responded with messages of the type $\langle(resp, seq_p), (v, s_O, S_C)\rangle$, where (v, s_O, S_C) correspond to the replica updated values of its triplet (v_p, obj_p, T_p) (lines 8–9).

Whenever a process (client or replica) p receives a new request, response or broadcast of the type $\langle msgType, (v, s_O, S_C)\rangle$, it updates its commit estimate and object candidate by joining its current values with the one received in the message. It also merge its set of input configurations T_p with the received input configurations, but the values superseded by the updated commit estimate are trimmed off T_p (lines 18–20). For replicas, they also send a response containing the updated triplet (v_p, obj_p, T_p) to the sender of the request (line 17).

If responses from quorums of all queried configurations are received and no response contained a *new*, not yet known, input configuration or a greater object state, then the couple formed by obj_p and the join the commit estimate configuration with all input configurations $\bigsqcup^{\mathcal{C}}(\{v.C\} \cup T_p)$ is broacasted and returned as the new learnt state (lines 12-14). Otherwise, clients proceed to a new round.

To ensure wait-freedom, we integrate a helping mechanism simply consisting in having clients adopt their committed state estimate (line 15). But, to know when a committed state is great enough to be returned, clients must first complete a communication round without interference from reconfigurations (line 11). After such a round, the join of all known states, stored in *learnLB*, can safely be used as lower bound to return a committed value.

Correctness. Let us first show that elements of the type $(v, s_O, S_C) \in \mathcal{L} \times \mathcal{O} \times 2^{\mathcal{C}}$ in which we have that $\forall u \in S_C, u \notin^{\mathcal{C}} v.C$ admits a partial order $\sqsubseteq^*$ defined as follows:

$$(v, s_O, S_C) \sqsubseteq^* (v', s'_O, S'_C) \Leftrightarrow v \sqsubseteq v' \wedge s_O \sqsubseteq^{\mathcal{O}} s'_O \wedge \{u \in S_C | u \notin^{\mathcal{C}} v'.C\} \subseteq S'_C.$$

Note that, since $\sqsubseteq$ and $\sqsubseteq^{\mathcal{O}}$ are partial orders, the reflexivity and transitivity properties are verified if they are verified by the relation $\{u \in S_C | u \notin^{\mathcal{C}} v'.C\} \subseteq S'_C$. Hence, the symmetry property is trivially verified as for any property $\mathcal{P}$, we have $\{u \in S_C | \mathcal{P}(u)\} \subseteq S_C$. For transitivity, $(v, s_O, S_C) \sqsubseteq^* (v', s'_O, S'_C)$ and $(v', s'_O, S'_C) \sqsubseteq^* (v'', s''_O, S''_C)$ implies that:

$$\{u \in S_C | u \notin^{\mathcal{C}} v''.C\} \subseteq \{u \in \{w \in S_C | w \notin^{\mathcal{C}} v'.C\} | u \notin^{\mathcal{C}} v''.C\} \subseteq \{u \in S'_C | u \notin^{\mathcal{C}} v''.C\} \subseteq S''_C.$$

Hence that $(v, s_O, S_C) \sqsubseteq^* (v'', s''_O, S''_C)$. For antisymmetry, given $(v, s_O, S_C) \sqsubseteq^* (v', s'_O, S'_C)$ and $(v', s'_O, S'_C) \sqsubseteq^* (v, s_O, S_C)$, the relations $\sqsubseteq$ and $\sqsubseteq^{\mathcal{C}}$ implies that $v = v'$ and $s_O = s'_O$. But as by assumption $\forall u \in S_C, u \notin^{\mathcal{C}} v.C$, we have $S_C = \{u \in S_C | u \notin^{\mathcal{C}} v.C\}$. But since $v = v'$ then $S_C = \{u \in S_C | u \notin^{\mathcal{C}} v'.C\}$, we obtain that $S_C \subseteq S'_C$. Likewise, we have $S'_C \subseteq S_C$, and thus, we obtain that $S_C = S'_C$, completing the verification of the antisymmetry property.

Intuitively, the set of elements $(v, s_O, S_C) \in \mathcal{L} \times \mathcal{O} \times 2^{\mathcal{C}}$, in which we have that $\forall u \in S_C, u \notin^{\mathcal{C}} v.C$, equipped with the partial order $\sqsubseteq^*$ is a join semi-lattice in which the procedure *updateState* replaces the triple (v_p, obj_p, T_p) with a join of itself and the procedure argument. But, we will only prove that the procedure *updateState* replace (v_p, obj_p, T_p) with an upper bound of itself and the procedure argument (v, s_O, S_C):

▸ **Lemma 1.** *Let* $(v_p^{old}, obj_p^{old}, T_p^{old})$ *and* $(v_p^{new}, obj_p^{new}, T_p^{new})$ *be the value of* (v_p, obj_p, T_p) *respectively before and after an execution of the* updateState *procedure with argument* (v, s_O, S_C)*, then, we have:*

$$(v_p^{old}, obj_p^{old}, T_p^{old}) \sqsubseteq^* (v_p^{new}, obj_p^{new}, T_p^{new}) \wedge (v, s_O, S_C) \sqsubseteq^* (v_p^{new}, obj_p^{new}, T_p^{new}).$$

Proof. Let us first note that we can rewrite the operation as follows:

- Line 18: $v_p^{new} = v_p^{old} \sqcup v$

Local variables:
seq_p, initially 0 { *The number of issued requests* }
v_p, initially (O_0, C_0) { *The last learnt state* }
T_p, initially $\varnothing$ { *The set of proposed configuration states* }
obj_p, initially O_0 { *The candidate object state* }

operation *propose*$(prop)$ { *Propose a new state prop* }
1 $updateState(v_p, prop.O, \{prop.C\})$
2 $learnLB := \bot$
3 **while** *true* **do**
4 $seq_p := seq_p + 1$
5 $oldCommit := v_p$ { *Archive commit estimate* }
6 $oldCandidates := (obj_p, T_p)$ { *Archive candidate states* }
7 $V := \{\bigsqcup^{\mathcal{C}}(\{v_p.C\} \cup S) \mid S \subseteq T_p\}$ { *Queried configurations* }
8 send $\langle (REQ, seq_p), (v_p, obj_p, T_p)\rangle$ to $\bigcup_{u \in V} members(u)$
9 **wait until** $oldCommit.C \neq v_p.C$ or $\forall u \in V$, received responses of the type $\langle (RESP, seq_p), \_\rangle$ from some $Q \in quorums(u)$
10 **if** $oldCommit.C = v_p.C \wedge oldCandidates = (\_, T_p)$ **then** { *Stable configurations* }
11 **if** $learnLB = \bot$ **then** $learnLB = (obj_p, \bigsqcup^{\mathcal{C}}(\{v_p.C\} \cup T_p)$
12 **if** $oldCandidates = (obj_p, \_)$ **then** { *No greater object received* }
13 *broadcast* $\langle COMMIT, ((obj_p, \bigsqcup^{\mathcal{C}}(\{v_p.C\} \cup T_p), obj_p, \varnothing)\rangle$
14 **return** $(obj_p, \bigsqcup^{\mathcal{C}}(\{v_p.C\} \cup T_p))$
15 **if** $learnLB \neq \bot \wedge learnLB \sqsubseteq v_p$ **then return** v_p { *Adopt learnt state* }

upon *receive* $\langle msgType, msgContent\rangle$ from process q
16 $updateState(msgContent)$ { *Update tracked states* }
17 **if** $msgType = (REQ, seq)$ **then** send $\langle (RESP, seq), (v_p, obj_p, T_p)\rangle$ to q

procedure *updateState*(v, s_O, S_C) { *Merge tracked states* }
18 $v_p := v_p \sqcup v$ { *Update the commit estimate* }
19 $obj_p := obj_p \sqcup^{\mathcal{O}} s_O$ { *Update the object candidate* }
20 $T_p := \{u \in (T_p \cup S_C) \mid u \not\sqsplit^{\mathcal{C}} v_p.C\}$ { *Update and trim input candidates* }

Figure 2 Reconfigurable universal construction: code for process p.

44 - Line 19: $obj_p^{new} = obj_p^{old} \sqcup s_O$
45 - Line 20: $T_p^{new} = \{u \in (T_p^{old} \cup S_C) | u \not\sqsplit^{\mathcal{C}} (v_p^{old} \sqcup^{\mathcal{C}} v).C\}$

46 Hence, the use of $(v_p^{old}, obj_p^{old}, T_p^{old})$ and (v, s_O, S_C) are symmetrical. Moreover, it is trivial
47 to check that, w.l.o.g., $(v, s_O, S_C) \sqsubseteq^* (v_p^{new}, obj_p^{new}, T_p^{new})$. Indeed, $v \sqsubseteq v_p^{old} \sqcup v$, $s_O \sqsubseteq^{\mathcal{O}}$
48 $obj_p^{old} \sqcup^{\mathcal{O}} s_O$ and $\{u \in S_C | u \not\sqsplit^{\mathcal{C}} v_p^{new}.C\} \subseteq \{u \in (T_p^{old} \cup S_C) | u \not\sqsplit^{\mathcal{C}} (v_p^{old} \sqcup^{\mathcal{C}} v).C\} = T_p^{new}$. ◂

49 Note that it is also trivial to check that initially we have $\forall u \in T_p, u \not\sqsplit^{\mathcal{C}} v_p.C$ as $T_p = \varnothing$
50 and that it remains true after a complete execution of the *updateState* procedure as T_p is
51 taken as the set of elements of $(T_p^{old} \cup S_C)$ satisfying this condition.
52 Let us now check that $\sqsubseteq^*$ is a refinement of the order $\sqsubseteq$ for the projection *decide*()
53 defined such that $decide(v, s_O, S_C) = (s_O, \bigsqcup^{\mathcal{C}}(\{v.C\} \cup S_C))$. Formally:

54 ▸ **Lemma 2.** $(v, s_O, S_C) \sqsubseteq^* (v', s'_O, S'_C) \implies decide(v, s_O, S_C) \sqsubseteq decide(v', s'_O, S'_C)$.

55 **Proof.** This result follows directly from the definition of $\sqsubseteq^*$. Indeed, as $(v, s_O, S_C) \sqsubseteq^*$
56 (v', s'_O, S'_C), we have $s_O \sqsubseteq^{\mathcal{O}} s'_O$. Moreover, we have $\{u \in S_C | u \not\sqsplit^{\mathcal{C}} v'.C\} \subseteq S'_C$. Hence we

have $\bigsqcup^{\mathcal{C}}(\{v'.C\} \cup S_C) \sqsubseteq^{\mathcal{C}} \bigsqcup^{\mathcal{C}}(\{v'.C\} \cup S'_C)$. But, as moreover we have $v \sqsubseteq v'$, we obtain that $\bigsqcup^{\mathcal{C}}(\{v.C\} \cup S_C) \sqsubseteq^{\mathcal{C}} \bigsqcup^{\mathcal{C}}(\{v'.C\} \cup S'_C)$. ◂

We are now going to show the main technical result required for the proof of correctness of Algorithm 2. Consider any run of the algorithm in Figure 2. Let s be any state committed in the considered run. Let $p(s)$ denote *the first* client that committed s in line 13. Let $V(s)$, $v(s)$, $obj(s)$ and $T(s)$ denote the value of respectively the variables V, $v_{p(s)}$, $obj_{p(s)}$ and $T_{p(s)}$ at the moment when $p(s)$ commited s in line 13. Note that, as $p(s)$ passed the tests in lines 10 and 12, $v_{p(s)}.C$, $obj_{p(s)}$ and $T_{p(s)}$ must have remained unchanged and equal to respectively $v(s).C$, $obj(s)$ and $T(s)$ since the last computation of V in line 7. In particular, we have $V(s) = \{\bigsqcup^{\mathcal{C}}(\{v(s).C\} \cup S) \mid S \subseteq T(s)\}$.

Let G be the graph whose vertices are all committed states plus $s_0 = (O_0, C_0)$ and whose edges are defined as follows:

$$s \to s' \;\Leftrightarrow\; s \sqsubsetneq s' \wedge s.C \in V(s').$$

Let us now show that G is connected, i.e., there exists a path between any couple of vertices in G:

▸ **Lemma 3.** *The graph G is connected.*

Proof. As $\sqsubseteq$ is a partial order, G is acyclic. Let s be any committed state, we have $v(s).C \in V(s)$ as $v(s).C$ is the value of $v_{p(s)}.C$ used in the computation of $V(s)$ in line 7. Hence, as $v(s) \sqsubseteq s$ since $s = \bigsqcup^{\mathcal{C}}(\{v(s)\} \cup T(s))$ and as $v(s) \neq s$ since $p(s)$ is the first process to commit s, any committed state admits a predecessor in G. Thus, the only source of G is s_0.

Let us show that G is connected by contradiction. Hence, let us assume that we can select s and s', a *minimal* (w.r.t. $\sqsubseteq$) pair of vertices of G that are not connected via a path, i.e., for all couple of vertices $(t, t') \neq (s, s')$ such that $t \sqsubseteq s$ and $t' \sqsubseteq s'$, there is path from t to t' or from t' to t in G.

Let us first show that s and s' share the same set of ancestors in G. Indeed, consider an ancestor u of s in G. As $u \sqsubsetneq s$ and as (s, s') is chosen minimal, there exists a path from u to s' or from s' to u. There is no path from s' to u as it would imply a path from s' to s. Hence, u is an ancestor of s'. By symmetry between s and s', we get that s and s' share the same set of ancestors in G. Let $\bar{s}$ be a locally maximal (w.r.t. $\sqsubseteq$) ancestor of s and s'. As there are no ancestors of s and s' greater than $\bar{s}$, the paths from $\bar{s}$ to s and s' are edges, i.e.:

- $\bar{s}.C \in V(s) \wedge \bar{s}.C \in V(s') \implies \bar{s}.C \in V(s) \cap V(s')$,
- $\bar{s} \sqsubsetneq s \wedge \bar{s} \sqsubsetneq s'$.

Let us now look back at the algorithm to show that a path must exists from s to s' or from s' to s. By the algorithm, as $\bar{s}.C \in V(s) \cap V(s')$, in the last round of requests before committing s (resp. s'), $p(s)$ (resp. $p(s')$) sent a request to all processes in $\bar{s}.C$. As, in their last round, $p(s)$ and $p(s')$ passed the test of line 10, they received responses from replicas of $\bar{s}.C$ forming *quorums* in $\bar{s}.C$, hence, as quorums intersect, from a common process $r \in \bar{s}.C$. Let us assume, w.l.o.g, that, for their last round of requests, r responded to $p(s)$ before responding to $p(s')$.

Recall that, as $p(s)$ passed the tests in lines 10 and 12, the values of $v_p.C$, obj_p and T_p did not change in the last round. Hence the content of the request sent to r by $p(s)$ is equal to $((v_O, v(s).C), obj(s), T(s))$, with v_O some arbitrary value. By Lemma 1, after r responsed to $p(s)$, (v_r, obj_r, T_r) must become and remain greater or equal to (w.r.t. $\sqsubseteq^*$) the message content $((v_O, v(s).C), obj(s), T(s))$. Hence, the latter response to $p(s)$ by r must contain a greater or equal content, and $(v_{p(s')}, obj_{p(s')}, T_{p(s')})$ becomes are remains greater or equal to $((v_O, v(s).C), obj(s), T(s))$, thus $((v_O, v(s).C), obj(s), T(s)) \sqsubseteq^* (v(s'), obj(s'), T(s'))$.

By Lemma 2, $s = decide((v_O, v(s).C), obj(s), T(s)) \sqsubseteq decide(v(s'), obj(s'), T(s')) = s'$.
As $v(s')$ is an ancestor of s', it is an ancestor of s, so $v(s).C \sqsubseteq^{\mathcal{C}} v(s').C \sqsubseteq^{\mathcal{C}} s.C$. Thus:

$$s.C = \overset{\mathcal{C}}{\bigsqcup}(\{v(s).C\} \cup T(s)) \sqsubseteq^{\mathcal{C}} \overset{\mathcal{C}}{\bigsqcup}(\{v(s').C\} \cup T(s)) \sqsubseteq^{\mathcal{C}} \overset{\mathcal{C}}{\bigsqcup}(\{s.C\} \cup T(s)) = s.C.$$

So $s.C = \bigsqcup^{\mathcal{C}}(\{v(s').C\} \cup T(s))$, and hence, $s.C = \bigsqcup^{\mathcal{C}}(\{v(s').C\} \cup \{u \in T(s), u \not\sqsubseteq^{\mathcal{C}} v(s').C\})$. From $((v_O, v(s).C), obj(s), T(s)) \sqsubseteq^* (v(s'), obj(s'), T(s'))$, we get that $\{u \in T(s), u \not\sqsubseteq^{\mathcal{C}} v(s').C\} \subseteq T(s')$, and therefore, we obtain that:

$$s.C = \overset{\mathcal{C}}{\bigsqcup}(\{v(s').C\} \cup \{u \in T(s), u \not\sqsubseteq^{\mathcal{C}} v(s').C\}) \in \{\overset{\mathcal{C}}{\bigsqcup}(\{v(s').C\} \cup S) \mid S \subseteq T(s')\} = V(s').$$

We have shown that $s \sqsubseteq s'$ and that $s.C \in V(s')$ and therefore that there is an edge, hence a path, from s to s' in G — A contradiction. ◂

We now have all the main ingredients to show the correctness of Algorithm 2.

▸ **Theorem 4.** *The algorithm in Figure 2 implements reconfigurable lattice agreement.*

Proof. For the **Consistency** property, Lemma 3 says that G is connected, and hence that all committed values are totally ordered, thus, that all learnt states are totally ordered.

From Lemma 3, we can also infer that $\forall s, s' \in G$, if $s \to s'$, then the first *propose* procedure returning s cannot preceed the first procedure returning s'. Indeed, In their last round of requests $p(s)$ ans $p(s')$ both queried $s'.C$, as $s'.C \in V(s)$ ans $s'.C \in V(s')$, and received responses from intersecting quorums, hence from a common process r. As shown in the proof of Lemma 3, this implies that the value committed by the first client r responded to is smaller than the other. Hence the procedure associated with s cannot preceed the procedure associated with s'. The same argument also holds for any other *propose* procedure committing s. hence, a client returning in line 14, return a state greater than all previously committed states, hence all previously learnt states.

For a process returning in line 15, to show that learnt states are greater than any preceeding learnt states, it is sufficient to check that *LearnLB* is greater than all. The selecte state *LearnLB* is not a committed state as the value of obj_p may have changed during the round of requests. But we can say that it is *semi-committed* as configurations did not change. This part is the most important as it is the property used in Lemma 3 to show that the client communicating latter with the common process r get a greater *decide*() state than the one committed by the first. Intuitively, this is sufficient to add semi-committed to the graph and show that there are path from semi-committed states to all smaller committed states, and hence that it is large enough to be greater than all previously committed, and hence, learnt states.

For the **Validity** property, we have shown that clients return states greater than all previously learnt states. By a trivial induction, as a committed state is a join of input states and committed states, it is easy to check that committed states, and hence learnt states, are joins of the initial state and input states. Moreover, as triples (v_p, obj_p, T_p) becomes and remains greater after the execution of line 1, then clients commit and set *LearnLB* to states greater than the procedure proposal. Hence returned states are greater than the procedure proposal. Therefore the **Validity** property is satisfied.

To prove the **Reconfigurable-Liveness** property, consider a run in which only finitely many distinct *configurations* are proposed. Hence, there exists a greatest learnt configuration state C_f. By the properties of the reliable-broadcast mechanism (line 13), eventually all

correct processes will receive a commit message including C_f. Hence, eventually, all correct processes will have $v_p.C = C_f$.

Assuming **configuration availability**, we have that every join of proposed configurations that are not yet superseded must have an available quorum. Thus, eventually, every configuration $u.C$ queried by correct processes are available. Therefore, correct processes cannot be blocked forever waiting in line 9 and, thus, has to perform infinitely many iterations of the while loop. Moreover, since configurations eventually no new configuration is discovered, all correct processes will eventually always pass the test in line 10 and therefore set a state for *learnLB*. In a round of requests after setting *learnLB* based on the triple (v_l, obj_l, T_l), the triple (v_r, obj_r, T_r) in all replicas from a quorum of C_f must become and remain greater (w.r.t $\sqsubseteq^*$) than (v_l, obj_l, T_l).

Now, let us assume that a correct process p never terminates, thus, it must observe greater object candidate at each round. This implies that infinitely many *propose* procedures are initiated, hence that a process commits infinitely may states. A committed state must be computed based on a triple (v_p, obj_p, T_p) greater than thoses in all received messages, in particular thoses from a quorum in C_f which must eventually be greater than (v_l, obj_l, T_l). Hence, evantally a committed state greater than *learnLB* is broadcasted, and this value is adopted and returned by p after receiving it — A contradiction. ◂

6 Reconfigurable objects

In this section, we use our reconfigurable lattice agreement (RLA) abstraction to construct an interval-linearizable reconfigurable implementation of any L-ADT L.

6.1 Defining and implementing reconfigurable L-ADTs

Let us consider two L-ADTs, an *object* L-ADT $L^{\mathcal{O}} = (A^{\mathcal{O}}, B^{\mathcal{O}}, (\mathcal{O}, \sqsubseteq^{\mathcal{O}}, \sqcup^{\mathcal{O}}), O_0, \tau^{\mathcal{O}}, \delta^{\mathcal{O}})$ and a *configuration* L-ADT $L^{\mathcal{C}} = (A^{\mathcal{C}}, B^{\mathcal{C}}, (\mathcal{C}, \sqsubseteq^{\mathcal{C}}, \sqcup^{\mathcal{C}}), C_0, \tau^{\mathcal{C}}, \delta^{\mathcal{C}})$ (Section 2).

The corresponding *reconfigurable L-ADT* implementation, defined on the composition $L = L^{\mathcal{O}} \times L^{\mathcal{C}}$, exports operations in $A^{\mathcal{O}} \times A^{\mathcal{C}}$. It must be interval-linearizable (respectively to $\mathcal{S}_L$) and ensure Reconfigurable Liveness (under the configuration availability assumption, Section 4).

In the reconfigurable implementation of L presented in Figure 3, whenever a process invokes an operation $a \in A^{\mathcal{O}}$, it proposes a state, $\tau(a, O_p)$—the result from applying a to the last learnt state (initially, C_0)—to RLA, updates O_p and returns the response $\delta(a, O_p)$ corresponding to the new learnt state. Similarly, to update the configuration, the process applies its operation to the last learnt configuration and proposes the resulting state to RLA.

▸ **Theorem 5.** *The algorithm in Figure 3 is a reconfigurable implementation of an L-ADT.*

Proof. Consider any execution of the algorithm in Figure 3.

By the Validity and Consistency properties of the underlying RLA abstraction, we can represent the states and operations of the execution as a sequence $z_0, I_1, z_1, \ldots, I_m, z_m$, where $\{z_1, \ldots, z_m\}$ is the set of learnt values, and each I_i, $i = 1, \ldots, m$, is a set of operations invoked in this execution, such that $z_i = \bigsqcup_{a \in I_i} \tau(a, z_{i-1})$.

A construction of the corresponding interval-sequential history is immediate. Consider an operation a that returned a value in the execution based on a learnt state z_i (line 2). Validity of RLA implies that $a \in I_j$ for some $j \leqslant i$. Thus, we can simply add a to set R_i. By repeating this procedure for every complete operation, we get a history $z_0, I_1, R_1, z_1, \ldots, I_m, R_m, z_m$

Shared: RLA, reconfigurable lattice agreement
Local:
  O_p, initially O_0 { *The last learnt object state* }
  C_p, initially C_0 { *The last learnt configuration* }
upon invocation of $a \in A^{\mathcal{O}}$ { *Object operation* }
1 $(O_p, C_p) := RLA.propose((\tau^{\mathcal{O}}(a, O_p), C_p))$
2 **return** $\delta^{\mathcal{A}}(a, O_p)$
upon invocation of $a \in A^{\mathcal{C}}$ { *Reconfiguration* }
3 $(O_p, C_p) := RLA.propose((\tau^{\mathcal{C}}(a, C_p), O_p))$
4 **return** $\delta^{\mathcal{C}}(a, C_p)$

Figure 3 Interval-linearizable implementation of L-ADT $L = L^{\mathcal{O}} \times L^{\mathcal{C}}$: code for process p.

complying with $\mathcal{S}_L$. By construction, the history also preserves the precedence relation of the original history.

Reconfigurable liveness of the implementation is implied by the properties of RLA (assuming reconfiguration availability). ◂

In the special case, when the L-ADT is *update-query*, the construction above produces a *linearizable* implementation:

▸ **Theorem 6.** *The algorithm in Figure 3 is a reconfigurable linearizable implementation of an update-query L-ADT.*

Proof. Consider any execution of the algorithm in Figure 3 and assume that L is update-query.

By Theorem 5, there exists a history $z_0, I_1, R_1, z_1, \ldots, I_m, R_m, z_m$ that complies with $\mathcal{S}_L$, the interval-sequential specification of L. We now construct a *sequential* history satisfying the *sequential* specification of L as follows:

- For every update u in the history, we match it with immediately succeeding matching response $\bot$ (remove the other response of u if any);
- For every response of a query q in the history we match it with an immediately preceding matching invocation of q (remove the other invocation of q if any);

As the updates of an L-ADT are commutative, the order in which we place them in the constructed sequential history S does not matter, and it is immediate that every response in S complies with τ and δ in a sequential history of L. ◂

6.2 L-ADT examples

We give three examples of L-ADTs that allow for interval-linearizable (Theorem 5) and linearizable (Theorem 6) reconfigurable implementations.

Max-register. The *max-register* sequential object defined on a totally ordered set V provides operations $writeMax(v)$, $v \in V$, returning a default value $\bot$, and $readMax()$ returning the largest value written so far (or $\bot$ if there are no preceding writes). We can define the type as an update-query L-ADT as follows:

$$MR_V = (writeMax(v)_{v \in V} \cup \{readMax\}, V \cup \{\bot\}, (V \cup \{\bot\}, \leqslant_V, max_V), \bot, \tau_{MR_V}, \delta_{MR_V}).$$

where $\leqslant_V$ is extended to $\bot$ with $\forall v \in V : \bot \leqslant_V v$, $\delta_{MR_V}(z, a) = z$ if $a = readMax$ and $\bot$ otherwise, and $\tau_{MR_V}(z, a) = max_V(z, v)$ if $a = writeMax(v)$ and z otherwise.

It is easy to see that $(V \cup \{\bot\}, \leqslant_V, max_V)$ is a join semi-lattice and the L-ADT MR_V satisfies the sequential *max-register* specification.

Set. The (add-only) *set* sequential object defined using a countable set V provides operations $addSet(v)$, $v \in V$, returning a default value $\bot$, and $readSet()$ returning the set of all values added so far (or $\varnothing$ if there are no preceding add operation). We can define the type as an update-query L-ADT as follows:

$$Set_V = (addSet(v)_{v \in V} \cup \{readSet\}, 2^V \cup \{\bot\}, (2^V, \subseteq, \cup), \varnothing, \tau_{Set_V}, \delta_{Set_V}).$$

where $\subseteq$ and $\cup$ are the usual operators on sets, $\delta_{Set_V}(z, a) = z$ if $a = readSet$ and $\bot$ otherwise, and $\tau_{Set_V}(z, a) = z \cup \{v\}$ if $a = addSet(v)$ and z otherwise.

It is easy to see that $(2^V, \subseteq, \cup)$ is a join semi-lattice and the L-ADT Set_V satisfies the sequential (add-only) *set* specification.

Abort flag. An *abort-flag* object stores a boolean flag that can only be raised from $\bot$ to $\top$. Formally, the LADT AF is defined as follows:

$$AF = \big(\{abort, check\}, \{\bot, \top\}, (\{\bot, \top\}, \sqsubseteq^{AF}, \sqcup^{AF}), \bot, \tau_{AF}, \delta_{AF}\big)$$

where $\bot \sqsubseteq^{AF} \top$, $\forall z \in \{\bot, \top\} : \top \sqcup^{AF} z = \top$, $\bot \sqcup^{AF} \bot = \bot$ $\tau_{AF}(z, abort) = \delta_{AF}(z, abort) = \top$, and where $\tau_{AF}(z, check) = \delta_{AF}(z, check) = z$.

Conflict detector. The *conflict-detector* abstraction [5] exports operation $check(v)$, $v \in V$ that may return *true* ("conflict"), or *false* ("no conflict"). The abstraction respects the following properties:

- If no two *check* operations have different inputs, then no operation can return *true*.
- If two *check* operations have different inputs, then they cannot both return *false*.

A conflict detector can be specified as an L-ADT defined as follows:

$$CD = \big(V, \{true, false\}, (V \times \{\top, \bot\}, \sqsubseteq^{CD}, \sqcup^{CD}), \bot, \tau_{CD}, \delta_{CD}\big)$$

where

- $\bot \sqsubseteq^{CD} \top$; $\forall v \in V$, $\bot \sqsubseteq^{CD} v$ and $v \sqsubseteq^{CD} \top$; $\forall v, v' \in V$, $v \neq v' \Rightarrow v \not\sqcup^{CD} v'$;
- $\tau_{CD}(z, v) = v$ if $z = \bot$ or $z = v$, and $\tau_{CD}(z, v) = \top$ otherwise;
- $\delta_{CD}(z, v) = true$ if $z = \top$ and *false* otherwise.

Also, we can see that $v \sqcup^Z v' = v'$ if $v = v'$ or $v = \bot$, and $\top$ otherwise.

▸ **Theorem 7.** *Any interval-linearizable implementation of CD is a conflict detector.*

Proof. Consider any execution of an interval-linearizable implementation of *CD*. Let S be the corresponding interval-sequential history.

For any two $check(v)$ and $check(v')$, $v \neq v'$, in S, the response to one of these operations must appear *after* the invocations of both of them. Hence, one of the outputs must be computed on a value greater than the join of the two proposals, equal to $\top$. Therefore, if both operations return, at least one of the them must return *true*.

The state used to compute the output must be a join of some invoked operations, hence operations can only return *true* if not all *check* operations share the same input. ◂

7 Applications

Many ADTs do not have commutative operations and, thus, do not belong to L-ADT. Moreover, many distributed programming abstractions do not have a sequential specification at all and, thus, cannot be defined as ADTs, needless to say as L-ADTs.

We show, however, that certain such objects can be *implemented from* L-ADT objects. As L-ADTs are naturally composable, the resulting implementations can be seen as using a single (composed) L-ADT object. By using a reconfigurable version of this L-ADT object, we obtain a reconfigurable version of the implemented type. In our implementations we omit talking about reconfigurations explicitly: to perform an operation on the configuration component of the system state, a process simply proposes it to the underlying RLA (see, e.g., Figure 3).

Our examples are atomic snapshots [1] and commit-adopt [17].

Atomic snapshots

An m-sized atomic-snapshot memory maintains an array of m positions and exports two operations, *update*(i, v), where $i \in \{1, \ldots, m\}$ is a location in the array and $v \in V$—the value to be written, that returns a predefined value ok and *snapshot*() that returns an m-vector of elements in V. Its sequential specification stipulates that every *snapshot*() operation returns a vector that contains, in each index $i \in \{1, \ldots, m\}$, the value of the last preceding *update* operation on the i^{th} position (or a predefined initial value, if there is no such *update* operation).

Registers using $MR_{\mathbb{N}\times V}$. We first consider the special case of a single register (1-sized atomic snapshot). We describe its implementation from a *max-register*, assuming that the set of values V is totally-ordered with relation $\leqslant^V$. Let $\leqslant^{reg}$ be a total order on $\mathbb{N} \times V$ (defined lexicographically, first on $\leqslant$ and then, in case of equality, on $\leqslant^V$). Let MR be a max-register defined on $\leqslant^{reg}$.

The idea is to associate each written value *val* with a *sequence number seq* and to store them in MR as a tuple (seq, val). To execute an operation *update*(v), the process first reads MR to get the "maximal" sequence number s written to MR so far. Then it writes $(s+1, v)$ back to MR. Notice that multiple processes may concurrently use $s + 1$ in their *update* operations. Ties are then broken by choosing the maximal value in the second component in the tuple. However, it is guaranteed that $s+1$ will be larger than the sequence number used by any *preceding update* operation. A *snapshot* operation simply reads MR and returns the value in the tuple.

Using any reconfigurable linearizable implementation of MR (Theorem 6), we obtain a reconfigurable implementation of an atomic (linearizable) register. Intuitively, all values returned by *snapshot* (read) operations on MR can be totally ordered based on the corresponding sequence numbers (ties broken using $\leqslant^V$), which gives the order of *reads* in the corresponding sequential history S.

Let *update*(v) be an operation such that (1) it writes tuple (s, v) to MR and (2) some read operation returned v after reading (s, v) in MR. We then insert this *update* operation in the sequential history S just before the first such read operation (if there are multiple such *update* operations, they can be inserted in a batch). Each remaining complete *update* operation is inserted either just before the first *update* in the history with a greater couple of sequence number and value or (if no such *update* exists) at the end of the history.

By construction, S is legal: every read returns the value of the last preceding write. Moreover, as only concurrent *updates* can use the same sequence number and the *snapshot* operations are ordered respecting the sequence numbers, S complies with the real-time precedence of the original history. We delegate the complete proof to the more general case of an m-sized snapshot.

Atomic snapshots. Our implementation of an m-sized atomic snapshot (described in

operation *update*(i, v) { *update register* i *with* v }
1 $(s, -) :=$ *MRset*$[i]$.readMax
2 *MRset*$[i]$.writeMax$((s + 1, v))$

operation *snapshot*()
3 $r :=$ *MRset.readAll*
4 **return** *snap* **with** $\forall i \in \{1, \ldots, m\}, r[i] = (-, snap[i])$

Figure 4 Simulation of an m-component atomic snapshot using an L-ADT.

Figure 4) is a straightforward generalization of the register implementation above. Consider the L-ADT defined as the product of m max-register L-ADTs. In particular, the partial order of the L-ADT is the product of m (total) orders $\leqslant^{snap}$: $\leqslant^{reg_1} \times \cdots \times \leqslant^{reg_m}$.

We also enrich the interface of the type with a new query operation *readAll* that returns the vector of m values found in the m max-register components. Notice that the resulting type is still an update-query L-ADT, as its (per-component) updates are commutative.

By Theorem 6, we can use a reconfigurable linearizable implementation of this type, let us denote it by *MRset*.

Now to execute *update*(v, i) on the implemented atomic snapshot, a process performs a read on the i^{th} component of *MRset* to get sequence number s of the returned tuple and perfroms *writeMax*$((s + 1, v))$ on the i^{th} component. To execute a snapshot, the process performs *readAll* on MR and returns the array of the second elements in the tuples of the returned array.

Similarly to the case of a single register, the results of all *snapshot* operations can be totally ordered using the $\leqslant^{snap}$ order on the vectors returned by the corresponding *readAll* calls. Placing the matching *update* operation accordingly, we get an equivalent sequential that respects the specification of atomic snapshot.

▸ **Theorem 8.** *Algorithm in Figure 4 implements an m-component MWMR atomic snapshot object.*

The Commit-Adopt Abstraction

Let us take a more elaborated example, the commit-adopt abstraction [17]. It is defined through a single operation *propose*(v), where v belongs to some input domain V. The operation returns a couple $(flag, v)$ with $v \in V$ and $flag \in \{commit, adopt\}$, so that the following conditions are satisfied:

- **Validity:** If a process returns $(\_, v)$, then v is the input of some process.
- **Convergence:** If all inputs are v, then all outputs are $(commit, v)$.
- **Agreement:** If a process returns $(commit, v)$, then all outputs must be of type $(\_, v)$.

We assume here that V, the set of values that can be proposed to the commit-adopt abstraction, is totally ordered. The assumption can be relaxed at the cost of a slighly more complicated algorithm.

Our implementation of (reconfigurable) commit-adopt uses a *conflict-detector* object *CD* (used to detect distinct proposals), a max-register MR_V (used to write non-conflicting proposals), and an *abort flag* object *AF*.

Our commit-adopt implementation is presented in Figure 5. In its *propose* operation, a process first accesses the *conflict-detector* object *CD* (line 1). Intuitively, the conflict detector makes sure that committing processes share a common proposal.

```
operation propose(v)
1   if CD.check(v) = false then          { check conflicts }
2       MR_V.writeMax(v)
3       if AF.check = ⊤ then return (adopt, v)          { adopt the input }
4       else return (commit, v)          { commit proposal }
5   else          { Try to abort in case of conflict }
6       AF.abort          { raise abort flag }
7       val := MR_V.readMax
8       if val = ⊥ then return (adopt, v)          { adopt the input }
9       else return (adopt, val)          { adopt the possibly committed value }
```

Figure 5 Commit-adopt implementation using L-ADTs.

If the object returns *false* (no conflict detected), the process writes its proposal in the max-register MR_V (line 2) and then checks the abort flag AF. If the check operation returns $\perp$, then the proposed value is returned with the *commit* flag (line 4). Otherwise, the smae value is returned with the *adopt* flag (line 3).

If a conflict is detected (CD returns *true*), then the process executes the *abort* operation on AF (line 6). Then the process reads the *max-register*. If a non-$\perp$ value is read (some value has been previously written to MR), the process adopts that value (line 9). Otherwise, the process adopts its own proposed value (line 8).

▸ **Theorem 9.** *Algorithm in Figure 5 implements commit-abort.*

Proof. The Validity property is trivially satisfied as processes return either their own proposal or the proposal of another process found in the max-register MR_V.

To prove Convergence, consider an execution in which all processes share the same input v. The conflict detector must return false to processes since it is accessed with a unique input. As no conflict is observed, no process could have called an *abort* operation on AF, and hence, the *check* operations on AF can only return $\perp$. Therefore all processes return with $(commit, v)$.

To prove Agreement, suppose, by contradiction, that the algorithm has an execution in which process p commits value v (line 4) and process q adopts or commits value $v' \neq v$ (in lines 4, 8 or 9).

We observe first that q cannot return in line 4, as otherwise the conflict detector would return *false* to p or q. For the same reason no value other than v could have been written to MT_V in this execution. Also, q must have completed line 6 before p checked AF in line 3, as otherwise p would not be able to commit v in line 4. Thus, q reads MR_V (line 7) *after* p has written v in it (line 2). Hence, q must have adopted the value read in MR_V (line 9), and this value must have been v—a contradiction. ◂

The Safe-Agreement Abstraction

Another popular shared-memory abstraction is *safe agreement* [10]. It is defined through a single operation $propose(v)$, $v \in V$ (we assume that V is totally ordered). The operation returns a value $v \in V$ or a special value $\perp \notin V$, so that the following conditions are satisfied:

- **Validity:** Every non-$\perp$ output has been previously proposed.
- **Agreement:** All non-$\perp$ outputs are identical.
- **Non-triviality:** If all participating processes return, then at least one returns a non-$\perp$ value.

```
operation propose(v)
1   In.addSet(id)            { enter the doorway }
2   if MR_V.readMax = ⊥ then MR_V.writeMax(v)        { write proposal if empty }
3   Out.addSet(id)           { exit the doorway }
4   outSet := Out.readSet
5   inSet := In.readSet
6   if inSet = outSet then return MR_V.readMax       { no process in doorway }
7   else return ⊥
```

Figure 6 Safe-agreement implementation using L-ADTs for process with identifier *id*.

Our implementation of safe agreement (Figure 6) uses two (add-only) sets denoted *In* and *Out* (Section 6) and a max-register MR_V.

The *propose* operation consists of two phases. In the first phase (lines 1-3) that we call the *doorway protocol*, the process add its identifier to *In*. Then the process reads MR_V. If $\perp$ is read, then the process writes its proposal to the max-register, and adds its identifier to the *Out* set.

In the second phase (lines 4-7), the process first reads *Out* and then—*In*. If the two sets match, then the process reads the max-register again and return the read value. Otherwise, the special value $\perp$ is returned.

Intuitively, the processes use the doorway protocol to ensure that only the first set of concurrently participating processes may write a value in the max-register. The second phase of the algorithm checks if there still can be processes poised to write to the max-register, and return the value of the max-register only if it is not the case.

▸ **Theorem 10.** *Algorithm in Figure 6 implements safe agreement.*

Proof. The Validity property is trivially satisfied, as any non-$\perp$ returned value must have been read in the max-register (line 6). As a process can read the max-register only after it has written its input in it (line 2), every such value must be an input value of some process.

To prove Agreement, consider, by contradiction, an execution in which two processes, p and q, return different non-$\perp$ values . Let p be the first process to read MR_V in line 6. Thus, the max-register MR_V has been written *after* it has been read (in line 6) by p and *before* it has been read (in line 6) by q. Let s be the process that performed the first such write.

Notice that before writing its input in MR_V, s must have read $\perp$ in it (line 2). Moreover, it must have executed line 2 *before* p has finished its doorway: otherwise s would find in MR_V the value written by p or an earlier written value. Thus, s has already added itself to the set *In* when p reads it in line 5. Furthermore, s is still in its doorway at the moment when p reads MR_V in line 6. In particular, s has not yet added itself to the set *Out* at that moment.

Thus, when p reaches line 6 its local variables *inSet* and *outSet* are not equal. Hence, p cannot return in line 6—a contradiction.

To prove Non-triviality, assume that all participating processes return and let p be the last process to write to the *Out* set. By that moment, all participating processes appear both in *In* and *Out*. Thus, p must return the value read in MR_V (line 6), which is non-$\perp$, as p has ensured before that (line 2). ◂

8 Related Work

Lattice agreement. Attiya et al. [8] introduced the (one-shot) lattice agreement abstraction and, in the shared-memory context, described a wait-free reduction of lattice agreement to atomic snapshot. Falerio et al. [15] introduced the long-lived version of lattice agreement (adopted in this paper) and described an asynchronous message-passing implementation of lattice agreement assuming a majority of correct processes, with $\mathcal{O}(n)$ time complexity (in terms of message delays) in a system of n processes. Our RLA implementation in Section 5 builds upon this algorithm.

CRDT. Conflict-free replicated data types (CRDT) were introduced by Shapiro et al. [28] for eventually synchronous replicated services. The types are defined using the language of join semi-lattices and assume that type operations are partitioned in updates and queries. Falerio et al. [15] describe a "universal" construction of a linearizable CRDT from lattice agreement. Skrzypczak et al. [29] argue that avoiding consensus in such constructions may bring performance gains. In this paper, we considered a more general class of types (L-ADT) that are "state-commutative" but not necessarily "update-query" and leveraged the recently introduced criterion of interval-linearizability [12] for *reconfigurable* implementations of L-ADTs using RLA.

Reconfiguration. *Passive reconfiguration* [7, 9] assumes that replicas enter and leave the system under an explicit *churn model*: if the churn assumptions are violated, consistency is not guaranteed. In the *active reconfiguration* model, processes explicitly propose configuration updates, e.g., sets of new process members. Early proposal, such as RAMBO [20] focused on read-write storage services and used consensus to ensure that the clients agree on the evolution of configurations.

Asynchronous reconfiguration. Dynastore [2] was the first solution emulating a reconfigurable atomic read/write register without consensus: clients can asynchronously propose incremental additions or removals to the system configuration. Since proposals commute, concurrent proposals are collected together without the need of deciding on a total order. Assuming n proposals, a Dynastore client might, in the worst case, go through 2^{n-1} candiate configurations before converging to a final one. Assuming a run with a total number of configurations m, complexity is $\mathcal{O}(min(mn, 2^n))$.

SmartMerge [23] allows for reconfiguring not only the system membership but also its quorum system, excluding possible undesirable configurations. SmartMerge brings an interesting idea of using an external reconfiguration service based on lattice agreement [15], which allows us to reduce the number of traversed configurations to $O(n)$. However, this solution assumes that this "reconfiguration lattice" is always available and non-reconfigurable (as we showed in this paper, lattice agreement is a powerful tool that can itself be used to implement a large variety of objects).

Gafni and Malkhi [18] proposed the *parsimonious speculative snapshot* task based on the commit-adopt abstraction [17]. Reconfiguration, built on top of the proposed abstraction, has complexity $\mathcal{O}(n^2)$: n for the traversal and n for the complexity of the parsimonious speculative snapshot implementation. Spiegelman, Keidar and Malkhi [30] improved this work by proposing a solution with time complexity $\mathcal{O}(n)$ by obtaining an amortized (per process) time complexity $\mathcal{O}(1)$ for speculative snapshots operations.

9 Concluding Remarks

To conclude, let us briefly discuss the complexity of our solution to the reconfiguration problem and overview how our solution could be further extended.

Round-trip complexity. The main complexity metric considered in the literature is the maximal number of communication round-trips needeed to complete a reconfiguration when n operations are concurrently proposed. In our solution, each time a round of requests is completed, a new input state was discovered to modify T_p or obj_p, hence we have at most n round-trips. Note that a round of requests might be interupted by receiving a greater committed state, at most n times as committed states are totally ordered joins of input states. The only other optimal solution with a linear round-trip complexity is from Spiegelman et al. [30]. In their solution the maximal number of round-trips is at least $4n$, that is twice than us. This has to do with the use of a shared memory simulation preventing to read and a write at the same time and preventing from sending requests to distinct configurations in parallel.

It is true that querying multiple configurations at the same time might increase the round-trip delay as we need to wait for more responses. Still, we believe that when the number of requests scales with a constant factor, this impact is negligeable.

Message complexity. The second metric that is studied in the literature is the number and size of the exchanges messages. In our protocol as in other solutions, messages are of linear size either for the distinct proposed configurations or the use of collect operations on the simulated memories.

The number of exchanged messages by our protocol may however greatly vary with the configuration object that is implemented. With at most k members per configuration, each client may send at most $k * 2^n$ messages per round as there is an exponential number of potential configuration to query. But this upper bound may be reached only if joins of proposed configurations do not share any replica. However, defining such configuration objects does not make much sense. In our example replicas may be added or removed, the one used in particular in most proposed solutions, clients may send at most $k + \Delta * n$ requests per round, where Δ the maximal number of replica added per proposal. In this case, the number of requests is comparable with k, the number of messages send to query a single configuration as done for solution based on a shared memory simulation.

An interesting question is whether we can construct a composite complexity metric that combines the number of messages a process sends and the time it takes to complete a *propose* operation. Indeed, one may try to find a trade-off between accessing few configurations sequentially versus accessing many configurations in parallel.

Optimizations. If the cost of querying many configurations in parallel outweigh the cost of contacting fewer configurations sequentially, one can proceed to a reconfigurable lattice agreement based on the methodology from [30]. Intuitively, it would consists in solving a generalized lattice agreement on the current configuration before switching the used configuration while using a carefully designed tracking mechanism of potentially used configurations.

A lighter modification to the RLA protocol may consists in leveraging timing constraints to wait for responses during a delay sufficient to obtain most responses, while waiting for responses from quorums only when no new information is received and an operation may return. Such modification may yield a great efficiency gain in practice as clients should be less constrained by slow responses while increasing the number of distinct inputs expected to discover per round.

Improvements can also be made for implemented objects when its lattice is well struc-

tured. A pertinent example is the fully ordered lattice states of max registers. For them, processes can directly return the state stored in *LearnLB* in line 11. Indeed, not returning a committed states might only violate the *consistency* property. But if states are totally ordered, then the *consistency* property is necessarily verified. Such a modification would yield to operations in a single round-trip when no reconfiguration occurs. Hence, it might be interesting to further investigate how the lattice structure might be leveraged in general.

References

1. Y. Afek, H. Attiya, D. Dolev, E. Gafni, M. Merritt, and N. Shavit. Atomic snapshots of shared memory. *J. ACM*, 40(4):873–890, 1993.
2. M. K. Aguilera, I. Keidar, D. Malkhi, and A. Shraer. Dynamic atomic storage without consensus. *J. ACM*, 58(2):7:1–7:32, 2011.
3. E. Alchieri, A. Bessani, F. Greve, and J. da Silva Fraga. Efficient and modular consensus-free reconfiguration for fault-tolerant storage. In *21st International Conference on Principles of Distributed Systems, OPODIS 2017, Lisbon, Portugal, December 18-20, 2017*, pages 26:1–26:17, 2017.
4. J. Aspnes, H. Attiya, and K. Censor. Max registers, counters, and monotone circuits. In *ACM Symposium on Principles of Distributed Computing*, PODC, pages 36–45, 2009.
5. J. Aspnes and F. Ellen. Tight bounds for adopt-commit objects. *Theory of Computing Systems*, 55(3):451–474, Oct 2014.
6. H. Attiya, A. Bar-Noy, and D. Dolev. Sharing memory robustly in message passing systems. *J. ACM*, 42(2):124–142, Jan. 1995.
7. H. Attiya, H. C. Chung, F. Ellen, S. Kumar, and J. L. Welch. Emulating a shared register in a system that never stops changing. *IEEE Trans. Parallel Distrib. Syst.*, 30(3):544–559, 2019.
8. H. Attiya, M. Herlihy, and O. Rachman. Atomic snapshots using lattice agreement. *Distributed Computing*, 8(3):121–132, 1995.
9. R. Baldoni, S. Bonomi, A. Kermarrec, and M. Raynal. Implementing a register in a dynamic distributed system. In *ICDCS*, pages 639–647, 2009.
10. P. Berman and A. A. Bharali. Quick atomic broadcast. pages 189–203. 93.
11. C. Cachin, R. Guerraoui, and L. Rodrigues. *Introduction to reliable and secure distributed programming*. Springer Science & Business Media, 2011.
12. A. Castañeda, S. Rajsbaum, and M. Raynal. Unifying concurrent objects and distributed tasks: Interval-linearizability. *J. ACM*, 65(6):45:1–45:42, 2018.
13. M. Castro and B. Liskov. Practical byzantine fault tolerance and proactive recovery. *ACM Transactions on Computer Systems (TOCS)*, 20(4):398–461, Nov. 2002.
14. G. V. Chockler, R. Guerraoui, I. Keidar, and M. Vukolic. Reliable distributed storage. *IEEE Computer*, 42(4):60–67, 2009.
15. J. Faleiro, S. Rajamani, K. Rajan, G. Ramalingam, and K. Vaswani. Generalized lattice agreement. In *PODC*, pages 125–134, 2012.
16. M. J. Fischer, N. A. Lynch, and M. S. Paterson. Impossibility of distributed consensus with one faulty process. *J. ACM*, 32(2):374–382, Apr. 1985.
17. E. Gafni. Round-by-round fault detectors (extended abstract): Unifying synchrony and asynchrony. In *PODC*, 1998.
18. E. Gafni and D. Malkhi. Elastic configuration maintenance via a parsimonious speculating snapshot solution. In *DISC*, pages 140–153, 2015.
19. D. K. Gifford. Weighted voting for replicated data. In *SOSP*, pages 150–162, 1979.
20. S. Gilbert, N. A. Lynch, and A. A. Shvartsman. Rambo: a robust, reconfigurable atomic memory service for dynamic networks. *Distributed Computing*, 23(4):225–272, 2010.

166 **21** M. Herlihy. Wait-free synchronization. *ACM Trans. Prog. Lang. Syst.*, 13(1):123–149, 1991.
167 **22** M. Herlihy and J. M. Wing. Linearizability: A correctness condition for concurrent objects.
168 *ACM Trans. Program. Lang. Syst.*, 12(3):463–492, 1990.
169 **23** L. Jehl, R. Vitenberg, and H. Meling. Smartmerge: A new approach to reconfiguration for
170 atomic storage. In *DISC*, pages 154–169, 2015.
171 **24** L. Lamport. The Part-Time parliament. *ACM Transactions on Computer Systems*,
172 16(2):133–169, May 1998.
173 **25** L. Lamport, D. Malkhi, and L. Zhou. Reconfiguring a state machine. *SIGACT News*,
174 41(1):63–73, 2010.
175 **26** M. Perrin. Concurrency and consistency. In *Distributed Systems*. Elsevier, 2017.
176 **27** F. B. Schneider. Implementing fault-tolerant services using the state machine approach: A
177 tutorial. *ACM Computing Surveys*, 22(4):299–319, Dec. 1990.
178 **28** M. Shapiro, N. M. Preguiça, C. Baquero, and M. Zawirski. Conflict-free replicated data
179 types. In *SSS*, pages 386–400, 2011.
180 **29** J. Skrzypczak, F. Schintke, and T. Schütt. Linearizable state machine replication of state-
181 based crdts without logs. *CoRR*, abs/1905.08733, 2019.
182 **30** A. Spiegelman, I. Keidar, and D. Malkhi. Dynamic reconfiguration: Abstraction and
183 optimal asynchronous solution. In *DISC*, pages 40:1–40:15, 2017.