

Support for Speculative Update Propagation and Mobility in Deno

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This paper presents the transactional framework of Deno, an object replication system specifically designed for use in mobile and weakly-connected environments. Deno uses weighted voting for availability and pair-wise, epidemic information flow for flexibility. This combination allows the protocols to operate with less than full connectivity, to easily adapt to changes in group membership, and to make few assumptions about the underlying network topology. These features are all crucial to providing effective support for mobile and weakly-connected platforms.

Deno has been implemented and runs on top of Linux and Windows NT/CE platforms. We use the Deno prototype to characterize the performance of two versions of Deno's protocol. The first version enables globally serializable execution of update transactions. The second supports a weaker consistency level that still guarantees transactionally consistent access to replicated data. The results show that our protocols either outperform or perform comparably to existing approaches, while achieving higher availability. Further, we show that the incremental cost of providing global serializability in this environment is low. Finally, we show that commit delays can be significantly decreased by allowing votes to be cast, and votes and updates to be disseminated, speculatively.

1. Introduction

This paper describes the design, implementation, and performance of Deno, a system that supports object replication in a transactional framework for mobile and weakly-connected environments. Deno's system model is illustrated in Figure 1. One or more clients connect to each peer server, which communicate through pair-wise information exchanges. The servers are not necessarily *ever* fully connected.

Deno's underlying protocols are based on an asynchronous protocol called *bounded weighted voting* [25, 26]. Asynchronous solutions [5, 11, 12, 24, 27] for managing replicated data have a number of advantages over traditional synchronous replication protocols in large-scale, mobile, and weakly-connected environments. They can operate with less than full connectivity, easily adapt to frequent changes in group membership, and make few demands on the underlying network topology. However, this functionality comes at a price. Asynchronous solutions are generally either slow and require reconciliations, or have lower availability because they rely on primary-copy schemes [32].

The focus of this paper is a new decentralized, asynchronous replica management protocol that addresses these concerns. The protocol retains the advantages of current asynchronous protocols, but generally performs better, has fewer connectivity requirements, and higher availability. No server ever needs to have complete knowledge of group membership, and a given server only needs to be in intermittent contact with at least one other server to take full part in the voting and commitment process. As such, the protocol is highly suited for environments with weak connectivity.

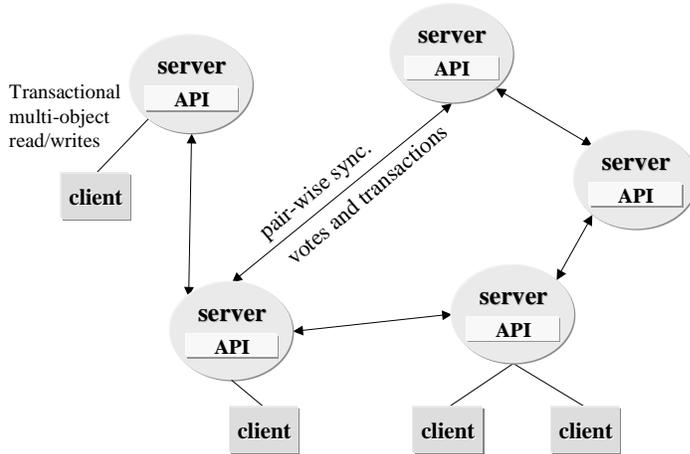


Figure 1: Basic Deno system model

Our prior work only defined consistency for single replicated objects. This paper extends bounded voting to include multi-item transactions and serializability. The voting protocol ensures mutual exclusion among conflicting transactions, guaranteeing that no two concurrent conflicting transactions can both commit. However, all transactions execute locally and no local or global deadlocks are possible.

The protocol’s strengths result from a novel combination of weighted voting and epidemic information flow [15], a process where information flows pairwise through a system like a disease passing from one host to the next. The protocol is completely decentralized. There is no primary server that “owns” an item or serializes the updates to that item (as in Bayou [34]). Any server can create new object replicas, and servers need only be able to communicate with a minimum of one other server at a time in order to make progress. Instead of *synchronously* assembling quorums, which has been extensively addressed by previous work (e.g., [18, 23, 35]), votes are cast and disseminated among system servers *asynchronously* through pair-wise, epidemic-style propagation. Any server can either commit or abort any transaction unilaterally, and all servers eventually reach the same decisions.

The use of voting allows the system to have higher availability than primary-copy protocols. The use of *weighted* voting allows implementations to improve performance by adapting currency distributions to site availabilities, update activity, or other relevant characteristics [13]. Each server has a specific amount of currency, and the total currency in the system is fixed at a known value. The advantage of a static total is that servers can determine when a plurality or majority of the votes have been accumulated *without complete knowledge of group membership*. This last attribute is key in dynamic, wide-area environments because it allows the protocol to operate in a completely decentralized fashion, eliminating performance bottlenecks and single points of failure.

The use of epidemic protocols divorces protocol requirements from communication requirements. First, an epidemic algorithm only requires protocol information to move throughout the system *eventually*. The lack of hard deadlines and connectivity requirements is ideally suited to mobile environments, where individual nodes are routinely disconnected. Second, epidemic protocols remove reliance on network topology. Synchronization partners in epidemic protocols are usually chosen randomly, eliminating the single point of failures that occur with more structured communication patterns, such as spanning trees.

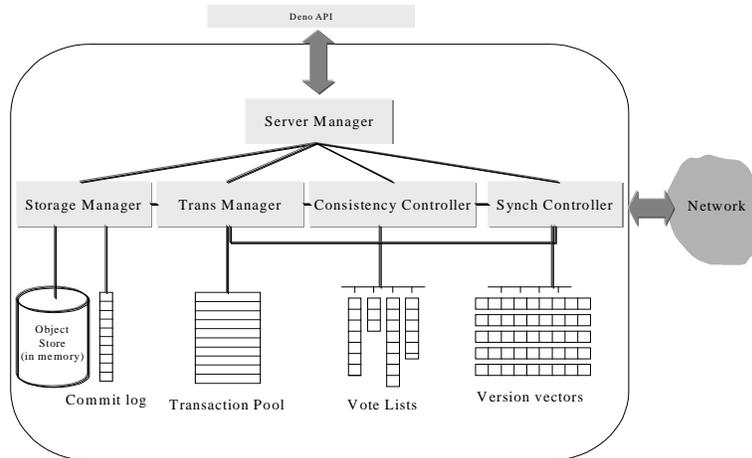


Figure 2: Basic Deno architecture

Our performance study is based on the Deno prototype. The basic Deno architecture has been implemented and runs on top of Linux/Win32/WinCE platforms. The performance data yielded three main findings. The overriding motivation for Deno’s protocols was to be able to make progress in weakly-connected environments. Protocols designed for such environments must make a number of tradeoffs that achieve availability at the possible expense of performance. Our first finding was that this performance impact was less than expected. On average, Deno servers learn of transaction commits just as fast as a much less available/reliable primary-copy protocol.

Our second finding was that support for global serializability is inexpensive in this environment. One of our protocols implements a form of weak consistency [10, 17], where update transactions are serializable and queries always access transactionally consistent database state. While this is sufficient for many applications, we also have a second variant that supports globally serializable executions, by providing a global commit order on all update transactions. Under both protocols, read-only transactions execute entirely at the local server, and do not require network communication.

Finally, we show that casting and disseminating votes speculatively can significantly improve the performance of protocols based on epidemic or similar communication mechanisms.

The rest of this paper is structured as follows. Section 2 describes the Deno architecture and prototype. Section 3 describes the base protocol, and Section 4 describes the extended version. Section 5 describes Deno’s support for mobility, and Section 6 presents the results of our performance study. Finally, Section 7 briefly describes related work, and Section 8 concludes.

2. Deno prototype

This section briefly describes the basic architecture of Deno object replication system. The overriding goal of the Deno project is to investigate replica consistency protocols for dis- and weakly-connected environments. We are therefore not motivated to build large and complicated interfaces to the object system. By the same token, we feel that lightweight interfaces are the appropriate choice for many applications, and that more complex services can be efficiently built on top of Deno services if needed. The basic Deno API supports operations for creating objects, creating and deleting object replicas, and performing reads and writes on the shared objects in a transactional framework.

2.1 Architecture

Figure 2 illustrates the basic Deno server architecture, consisting of the following components:

- The *Server Manager* is in charge of coordinating the activities of the various components. It handles client requests by implementing the basic Deno API.
- The *Consistency Controller* implements the decentralized voting protocols used by Deno. In particular, it maintains a *vote pool* that summarizes the votes known to the server.
- The *Synch Controller* is responsible for implementing efficient synchronization sessions with other Deno servers by maintaining *version vectors* that compactly summarize the events of interests from other servers. This component implements different synchronization policies that specify when and with whom to synchronize. In the current implementation, it implements a naïve policy that chooses synchronization partners randomly at regular intervals.
- The *Trans Manager* is mainly responsible for the local execution of transactions. It maintains a transaction pool that contains all active (i.e., non-obsolete) transactions known to the server.
- The *Storage Manager* provides access to the *object store* that stores the current committed versions of all replicated objects at the server. The object store is currently a simple in-memory database.

The prototype makes relatively few demands on the operating system and is therefore highly portable. The current prototype runs on top of Linux and WindowsNT/CE platforms. All communication is made on top of UDP/IP. Deno’s source consists of ~10,000 lines of multi-threaded C++ code.

3. Decentralized replication protocols

Before delving into the fine detail, we give a quick overview of the “life” of a Deno transaction (Figure 3). A transaction is submitted by a client to any server, which executes it locally. Upon completion, the transaction either blocks (if the local server has seen a conflicting transaction) or becomes a *candidate*, which means that the update can become visible to other servers. Candidates are voted on, and eventually either commit (if they corner a plurality of the system currency), or are aborted. We now describe Deno’s replication framework employed in detail.

3.1 Providing weak consistency: Base protocol

3.1.1 Transaction model

A transaction consists of a sequence of read and write operations on replicated data items. A transaction reads a set of *read items*, and updates a subset of the read items called *update items*. Our model assumes no blind-writes, i.e., data items are always read before being updated. Current values are tracked by associating a version number with each database item.

Each Deno server maintains an *active transaction list* that contains *active* transactions; i.e., transactions that are being executed. While a transaction is executing, it constructs a *transaction record* that summarizes the transaction’s execution state. A transaction record for a transaction t enumerates the read items of t (along with the version numbers of the items t recorded when it read the items), the update items of t , and the new values of the update items written by t . A transaction does not perform any locking before it performs an operation; it simply performs the operation and records it in its transaction record. Furthermore, update transactions do not perform *in-place* updates; i.e., a transaction does *not* update the local copy of the item stored in the database. Instead, it simply writes the new value in its transaction record. When a transaction *re-reads* an item it has already updated, the transaction reads the value it has most recently written. Likewise, when a transaction *re-writes* a new value, it overwrites the value it has previously written in its record.

The items in the local copy of the database are modified, and their version numbers incremented, only when update transactions commit. In other words, transactions only access committed values. Depending

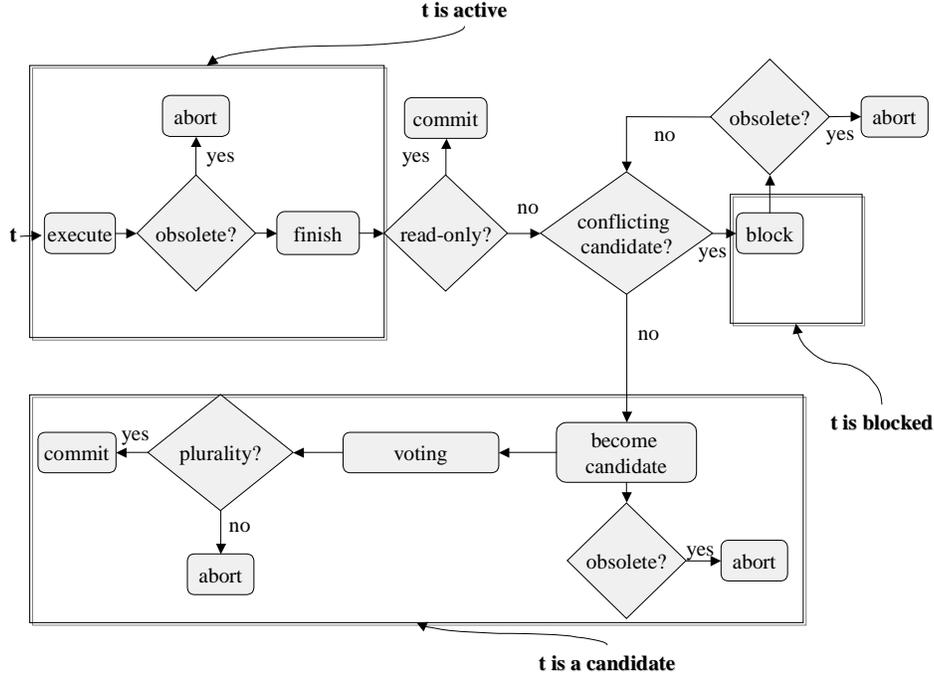


Figure 3: A transaction's life

on application semantics, however, this requirement can be relaxed, and transactions may be allowed to see new values written by uncommitted transactions. Such a model may especially be useful in facilitating disconnected operation, and has been investigated in the context of Bayou [34].

We distinguish between two types of transactions, *queries* (i.e., read-only transactions) and *update transactions*. Both types of transactions execute entirely locally. However, queries are more light-weight in that a query can commit without further processing immediately after it successfully finishes its execution. Update transactions, on the other hand, must participate in a distributed commitment process after finishing execution. When an active transaction successfully completes its execution, it takes one of the following two paths: (1) The transaction can either become a candidate transaction at its local server and participate in a distributed voting process that determines whether it commits or aborts; or (2) the transaction blocks and waits for the termination of other previous transactions before becoming a candidate. The blocked transactions are later considered for becoming candidates. In Section 4, we eliminate this restriction mainly in order exploit commutativity.

3.1.2 Voting

Define V_i as the set of all votes seen (the *vote set*) by Deno server s_i . A vote $v \in V_i$, is a 4-tuple $\langle voter, trans, curr, tstamp \rangle$ where:

- $v.voter$ denotes the server that casts the vote,
- $v.trans$ denotes the transaction the vote is cast for,
- $v.curr$ denotes the amount of currency $v.voter$ voted for $v.trans$ ¹,
- $v.tstamp$ is the value of the local $v.voter$'s local timestamp when $v.trans$ was created. A timestamp is a monotonically increasing local counter that is incremented each time the server casts a vote.

¹ Servers can dynamically exchange currencies in order to improve availability and performance. This issue, however, is outside the scope of this paper.

Notice that the combination of “*voter*” and “*trans*” fields uniquely identify a vote. In the following, we omit the other fields of a vote when they are not relevant to the discussion.

Let $t_i.read$ be the read items and $t_i.update$ be the update items of t_i . Furthermore, let $t_i.vers(d)$ denote the version of item d observed (and recorded) by transaction t_i when t_i first accessed d .

Definition 1 Two transactions, t_i and t_j , are said to *conflict* if (1) their common read items have the same version numbers, and (2) one of the transaction’s read items overlap with the other’s update items. More formally, t_i and t_j conflict if:

- (1) $t_i.vers(d) = t_j.vers(d), \forall d \in t_i.read \cap t_j.read$, and
- (2) $(t_i.read \cap t_j.update) \cup (t_i.update \cap t_j.read) \neq \emptyset$.

A server, s_i , votes for a transaction by creating a vote, v , assigning a currency value to v , and inserting it into V_i . The currency value for a vote can be set in two distinct ways based on the state of the vote set. Server s_i votes with its full currency for transaction t_i if it has not already voted for a conflicting candidate transaction. Such a vote is called a *yes* vote and is an indication of the support of the server for the corresponding transaction. Otherwise, s_i votes with 0.0 currency, in which case the vote is called a *no* vote.

We now describe the voting process from the perspective of a single server. Each Deno server s_i maintains the following major data structures:

1. a set of votes, V_i .
2. a list of *candidate* transactions, C_i . C_i consists of those update transactions that are known to s_i , have finished execution either locally or remotely, but have yet to be either committed or aborted at s_i .
3. a list of *blocked* transactions, B_i , consisting of locally completed transactions waiting to become candidates.
4. a commit log, which contains a list of committed transaction records.

A server may create a *vote* for a candidate or locally completed transaction that does not conflict with any other candidate transaction for which it has also voted. If the server votes for a blocked transaction, the transaction becomes a candidate transaction and is moved from the blocked list to the candidate list. Once made, votes may not be retracted. As explained below, a transaction t *commits* at s_i when it is guaranteed that no conflicting transaction can obtain more votes. Transactions can be committed even without knowledge of complete group membership because the total amount of currency in the system is *always* 1.0. The protocol guarantees that all servers eventually reach the same commit decisions.

Voting rule: Server s_i considers voting for a transaction in the following three cases:

1. *When s_i learns about a new candidate transaction t after synchronizing with another server;* it votes yes for t if it has not already voted for a conflicting transaction; Otherwise, s_i votes *no*.
2. *When s_i commits or aborts a candidate transaction;* it considers all transactions t in the blocked list (i.e., all transactions waiting to become candidates) in insertion order. For any such transaction that does not conflict with an existing candidate transaction; s_i votes *yes*.
3. *When s_i completes the execution of a local transaction t ;* if there is no candidate transaction that conflicts with t , s_i votes *yes* for t and inserts t into its candidate list, C_i . Otherwise, s_i blocks t and inserts t into its blocked list, B_i .

There are two important implications of the cases stated above. First, there cannot exist *yes* votes from the same server for conflicting transactions. Second, locally completed transactions are blocked until the termination of existing *conflicting* candidate transactions.

3.2 Update commitment

Given a server s_i , and its vote set V_i , we make the following definitions:

Definition 2 We define the sum of votes cast for a transaction t as: $votes(t) = \sum v.curr$,
where $v \in V_i$, and $v.trans=t$.

Definition 3 We define the *unknown votes* of a transaction t as: $unknown(t) = 1.0 - \sum s.curr$,
where s is a server who has already voted for t .

In other words, $unknown(t)$ is essentially the sum of currencies of those servers whose votes for transaction t are *not* available. We now define the commit rule that server s_i uses to decide which candidate transactions to terminate (i.e., commit or abort) on the basis of local information (i.e., votes that are locally available). The fundamental idea is to commit a transaction when it is guaranteed that no other conflicting transaction can gather more votes. Note that $server(t)$ is defined as the server on which transaction t was executed.

Commit Rule: A transaction $t \in C_i$ commits when: $\forall t' \in C_i$ such that t' and t conflict:

1. $votes(t) > votes(t') + unknown(t)$, or (*Plurality case*)
2. $votes(t) = votes(t') + unknown(t)$ and $server(t) < server(t')$ (*Tie-break case*)

The commit rule states that candidate transaction t can commit if it gathers the *plurality* of votes. Case 1 enforces mutual exclusion, as indicated before, by ensuring that no other conflicting transaction, which may or may not be known to server s_i , can gather more votes. Case 2 states that ties are broken using a simple deterministic comparison between the indices of the servers that created the transactions.

When a candidate transaction t commits at server s_i , s_i incorporates the effects of t into its database by installing the new values of the update items of t (available from t 's transaction record), and incrementing the version numbers of the local copies of those items. Finally, the transaction record of t is appended to the commit log. Note that servers must eventually garbage-collect their commit logs, as otherwise these logs will grow indefinitely. The candidate transactions and votes, on the other hand, do not need to be garbage-collected because they are discarded as soon as they become obsolete.

All active and candidate transactions whose read items are modified become *obsolete* and need to be aborted. More formally:

Abort Rule: An active, blocked or candidate transaction, $t \in C_i$, is aborted at s_i when there exists a data item d , s.t.:

1. $d \in t.read$, and
2. $s_i.vers(d) > t.vers(d)$.

where $vers(d)$ is the version of d observed by a transaction or server. Additionally, committing a transaction causes all votes cast for an obsolete transaction to be discarded, i.e., a vote $v \in V_i$ is discarded if $v.trans=t$ and t has become obsolete.

3.3 Synchronization

We now discuss how two Deno servers synchronize their states.² A pair-wise synchronization session essentially involves the propagation of (1) committed transactions, (2) candidate transactions, and (3) votes that are known to one server and unknown to the other.

In Deno, synchronization is controlled via version vectors [28]. In our model, each server s_i maintains an n -element vector, vv_i , where n is the number of servers, that describes the number of events of each

² This process is also called *anti-entropy* [15] in the terminology of epidemic algorithms.

other server *seen* by s_i . Element $vv_i[j]$ is a scalar count of the number of j 's events that have been seen at s_i . For example, if $vv_i = \{2, 3, 1\}$, then s_i has had two local events, has seen three events of s_2 , and has seen one event of s_3 's. There are three types of events: transaction commits, transaction promotions, and votes. A commit event is created whenever the local process commits a transaction, a promotion event is created whenever a transaction becomes a candidate on the server where it executed, and a vote event is created whenever a vote is cast.

In more detail, server s_i maintains a serial order, called *local ordering*, on all local commits, promotions and votes. We denote the j^{th} such event as e_i^j . As information about events is always propagated in local order, we can maintain the following invariant:

Synchronization invariant: If s_i 's version vector is vv_i , s_i has seen all events $e_j^1 \dots e_j^{vv_i[lj]}$, for all $j = 1 \dots n$.

Synchronization is then straightforward. For purposes of exposition, we assume a unidirectional *pull* synchronization, although other modes are possible [15, 26]. When s_i pulls information from s_j , the following actions take place:

1. Server s_i sends vv_i to s_j .
2. Server s_j responds with all events e_k^l s.t. $l > vv_i[k]$ and $l \leq vv_j[k]$, for all $k = 1 \dots n$.
3. Server s_i incorporates the new events in the same order that they originally occurred by:
 - a. processing new commitments, candidates, and votes,
 - b. applying the voting rule, the commit rule, and the abort rule for all relevant transactions,
 - c. updating vv_i to the pairwise maximum of vv_i and vv_j

For purposes of exposition, we assumed n -dimensional vectors in the above description. As we described in the introduction, however, we do not assume that the number of servers is known to any server. Our implementation uses a set representation for the version vector, i.e.:

$$vv_i = \{(j, c_j), (k, c_k) \dots\}$$

where each pair consists of a server id, j , and a count specifying the number of j 's events seen by s_i . The lack of a pair in vv_i describing some server k would be treated as an implicit pair $(k, 0)$, meaning that no events from that server have been seen.

3.4 Protocol illustration

Let $D(t)$ represent the set of read items of t , with the update items marked by a “*”. For the following two examples, assume transaction records and initial vote sets as follows:

$$D(t_1) = \{d_1, d_2^*\}, D(t_2) = \{d_1, d_2^*\}, D(t_3) = \{d_1, d_4^*\}, D(t_4) = \{d_2, d_3, d_4^*\}$$

$$V_1 = \{\langle s_1, t_1, 0.2 \rangle, \langle s_1, t_2, 0.0 \rangle, \langle s_1, t_3, 0.2 \rangle, \langle s_1, t_4, 0.0 \rangle,$$

$$\langle s_2, t_2, 0.2 \rangle,$$

$$\langle s_3, t_2, 0.25 \rangle, \langle s_3, t_3, 0.25 \rangle,$$

$$\langle s_4, t_2, 0.0 \rangle, \langle s_4, t_4, 0.25 \rangle\},$$

$$V_4 = \{\langle s_2, t_2, 0.2 \rangle,$$

$$\langle s_4, t_2, 0.0 \rangle, \langle s_4, t_4, 0.25 \rangle\}.$$

Let $c(t_i, t_j)$ mean that transaction t_i conflicts with t_j . Then, $c(t_1, t_2)$, $c(t_1, t_4)$, $c(t_2, t_4)$, and $c(t_3, t_4)$.

Example 1: Consider server s_1 . Using commit rule 1, s_1 commits t_2 because $votes(t_2) = 0.45$, $unknown(t_2) = 0.10$, and no conflicting transaction has a vote of more than 0.25, i.e. the maximum any conflicting transaction could gather is guaranteed to be less than t_2 's current votes. Thus, s_1 aborts the can-

didate transactions, t_1 and t_4 , and discards the votes that became obsolete at s_1 by the commitment of t_2 . Afterwards, $V_1 = \{ \langle s_1, t_3, 0.2 \rangle, \langle s_3, t_3, 0.25 \rangle \}$.

Example 2: Suppose that s_4 now pulls information from s_1 . The events propagated from s_1 to s_4 consist of the commit of t_2 , candidate t_3 , and votes $\langle s_1, t_3, 0.2 \rangle$ and $\langle s_3, t_3, 0.25 \rangle$. The arrival of the commit decision causes s_4 to commit t_2 , and to discard t_4 . At the end of the synchronization, $V_4 = \{ \langle s_1, t_3, 0.2 \rangle, \langle s_3, t_3, 0.25 \rangle \}$. At this point, s_4 votes *yes* for t_3 , and adds $\langle s_4, t_3, 0.25 \rangle$ into V_4 . It then applies the commit rule, and commits t_3 since $votes(t_3) = 0.70$. Server s_4 then discards all relevant votes. Finally, $V_4 = \{ \}$.

3.5 Correctness and consistency issues

We now discuss the consistency level provided by the base voting protocol. First, we present the following lemma:

Lemma 1 (*Update Consistency*) If an update transaction t commits at one server, then t eventually commits at all servers.

Proof (*Sketch*): Assume that transaction t_i committed at server s_i . Let $yes(t_i)$ denote the set of servers that voted *yes* for t_i . Now consider another server s_j and another transaction t_j that conflicts with t_i . If all the votes cast by the servers in $yes(t_i)$ are known at s_j , then s_j cannot commit t_j . Even if s_j may not know the votes cast by some of the servers in $yes(t_i)$, that amount will be reflected in $unknown(t_j)$, preventing t_j from committing at s_j . Therefore, s_j will eventually deduce the same outcome as s_i and commit t_i itself, or be told of the commitment of t_i by another server.

Lemma 2 (*Update Serializability*) The base voting protocol ensures global update serializability.

Proof (*Sketch*): Assume that the protocol generates a non-serializable global schedule involving update transactions. Then, by the previous lemma, there exists a cycle in the global serialization graph³ of the form $t_1 \rightarrow t_2 \rightarrow \dots \rightarrow t_n \rightarrow t_1$ where t_1, t_2, \dots, t_n are update transactions. Consider t_1 and t_2 . Since $t_1 \rightarrow t_2$, t_1 and t_2 must conflict on some data item, d . Suppose t_1 commits before t_2 at server s . Assume now that t_2 committed at s' before t_1 . We consider the three possible types of conflicts between t_1 and t_2 at s' :

1. *rw* (t_2 writes an item d which is then read by t_1): Since t_2 updated d when it committed at s' , the version number of d recorded by t_1 will be strictly smaller than the version number of the copy of d at the database of s' . This establishes t_1 as an obsolete transaction at s' and leads to t_1 being aborted.
2. *wr* (t_2 reads an item written by t_1): This case is the opposite of the previous case. This time, t_2 cannot commit at s , as it is based on a version of d that has already been updated by t_1 .
3. *ww* (t_2 writes an item written by t_1): This conflict type implies both *rw* and *wr* conflicts among t_1 and t_2 . It is, therefore, subsumed by the previous two cases (since we do not allow blind-writes).

We therefore conclude that t_1 must have committed before t_2 at all servers. A straightforward induction based on the transitivity of the conflict relation asserts that t_1 commits before t_n at all servers. This eliminates the possibility of a cycle in the serialization graph, thereby producing the contradiction that completes the proof.

So far, we have addressed only update transactions and showed that our protocol guarantees the serializable execution update transactions alone. The protocol thus prohibits serialization graph cycles that contain only update transactions. We now extend our discussion to include queries, and demonstrate that the protocol supports a weak form of consistency, which we define below:

Definition 4 (*Weak Consistency*): A query sees weak consistency if it serializes with respect to all update transactions, but possibly not with other queries [9, 10, 17].

³ A serialization graph [9] consists of nodes that represent transactions and directed edges that represent conflicting operations. A path from t_i to t_j indicates that t_i has to precede t_j in any equivalent serial ordering.

In weak consistency, each query observes a serial order of update transactions, which is not necessarily the same order observed by other queries. However, weak consistency does ensure that queries always observe transactionally-consistent database states. In other words, a query does not see partial effects of any update transaction. Weak consistency prohibits both *update transaction cycles* (i.e., cycles involving only update transactions), and *single-query cycles* (i.e., cycles involving a single query and one or more update transactions).

Lemma 3 (*Query-Transaction ordering*) Let q be a query and t be an update transaction that respectively reads and updates item d . The dependency $q \rightarrow t$ implies that q commits before t commits, and $t \rightarrow q$ implies that t commits before q commits, at the execution server of q .

Proof (*Sketch*): First consider $q \rightarrow t$. Query q reads d before t updates d . Query q must have committed before t committed. Otherwise, q must have been active when t committed, and the commitment of t would have aborted q (as q would have become obsolete). Now consider $t \rightarrow q$; q reads d after t updates d . In this case, q must have read d and committed after t since any update transaction (including t) installs its updates and commits atomically.

Theorem 1 (*Weak Consistency*) The base protocol described above provides weak consistency.

Proof (*Sketch*): Assume that there is a single-query cycle, involving query q and update transactions t_1, t_2, \dots, t_n , of the form $q \rightarrow t_1 \rightarrow t_2 \rightarrow \dots \rightarrow t_n \rightarrow q$. Consider $q \rightarrow t_1$. By Lemma 3, q must have committed before t_1 at the execution server of q , say s . By Lemma 1, t_1 commits before t_n at all servers. Therefore, q must have committed before t_n at s , prohibiting the single-query cycle assumed initially. Moreover, we know by Lemma 2 that there can not be any update transaction cycles. Therefore, we conclude that the protocol provides weak consistency.

4. Providing serializability: Extended protocol

The base protocol ensures that queries always access transactionally-consistent data, and that update transactions are globally serialized with respect to each other. However, as the following example illustrates, the base protocol does not serialize update transactions with respect to all queries:

Example 3: Consider update transactions t_1 and t_2 , where $D(t_1)=\{d_1^*\}$ and $D(t_2)=\{d_2^*\}$; and queries q_1 and q_2 , where $D(q_1)=D(q_2)=\{d_1, d_2\}$. Suppose server s_1 commits t_1, q_1 , and t_2 ; and server s_2 commits t_2, q_2 , and t_1 , in the order presented. This scenario is perfectly valid using the base protocol since t_1 and t_2 do not conflict and, thus, they are not necessarily serialized globally. The local serialization graphs at s_1 and s_2 will then be, $t_1 \rightarrow q_1 \rightarrow t_2$ and $t_2 \rightarrow q_2 \rightarrow t_1$, respectively. The global serialization graph, therefore, has a cycle.

This section extends the base protocol to provide strong consistency, which provides serializability to queries as well. We define strong consistency as follows:

Definition 5 (*Strong Consistency*) A query sees strong consistency if it is serialized with respect to both queries and update transactions. Strong consistency is characterized by an acyclic serialization graph, prohibiting both *update transaction cycles* and *multi-query cycles* (i.e., cycles involving multiple queries and one or more update transactions). This form of consistency guarantees globally-serializable executions [9, 10, 17].

The base protocol fails to provide strong consistency because non-conflicting update transactions are not necessarily globally serialized with respect to each other. We address this problem by forcing all update transactions to commit in the same order at all servers⁴ by providing mutual exclusion among *all* transac-

⁴ In fact, there are at least two other approaches to provide strong consistency. One approach is to include queries in the global voting process, which is clearly not desirable in our target environments. A second approach is to commit an update transaction

tions, rather than just among conflicting transactions as the base protocol does. We accomplish this by modifying the voting process such that each server votes *yes* for all candidate transactions (whether or not they conflict), but specifies a total order on all of its votes. The commit process is then restricted so that only the *top* transactions, which are the candidate transactions that come first in any server's ordering, are considered for commitment.

More specifically, the protocol works as follows. Instead of choosing among conflicting transactions, a server votes *yes* for all candidate transactions as soon as they are received. The result is that V_i contains a *yes* vote by s_i for each candidate transaction, differing only in the votes' timestamps. The timestamps impose a total ordering on all votes created by s_i . A transaction may be committed if it gains a plurality of the *top votes*, where a top vote is the earliest vote in the vote set from a specific server. More formally:

Definition 6 A vote $v \in V_i$ is said to be a *top vote* at server s_i if:

$$v.timestamp < v'.timestamp, \forall v' \in V_i \text{ s.t. } v.voter = v'.voter.$$

A candidate transaction $t \in C_i$ is said to be a *top transaction* at s_i if:

$$\exists v \in V_i \text{ s.t. } v.trans = t \text{ and } v \text{ is a top vote at } s_i.$$

Definition 7 We define the sum of votes cast for a *top* transaction t as:

$$votes(t) = \sum v.curr, \text{ where } v \in V_i \text{ s.t. } v.trans = t.$$

Definition 8 We define the unknown votes as:

$$unknown = 1.0 - \sum votes(t'), \text{ where } t' \text{ is a top transaction.}$$

Any server s_i may have up to n top votes and n top transactions, one of each for each of the n servers in the system. Notice the difference between Definition 3 and Definition 8. Definition 3 indicates that each transaction may have a different “*unknown*”. In the modified definition, the “*unknown*” value applies to the vote set of an entire server. We now state the modified commit rule a server s_i employs:

Commit Rule (Strong Consistency): A *top* transaction $t \in C_i$ commits when, $\forall t' \in C_i$ such that t' is a top transaction:

1. (*Plurality case*) $votes(t) > votes(t') + unknown$, or
 2. (*Tie-break case*) $votes(t) = votes(t') + unknown$ and $server(t) < server(t')$.
-

Aborts are handled as in Section 3.2. The following example illustrates how the extended protocol works:

Example 4: Consider server s_1 with the following transactions and votes as follows:

$$D(t_1) = \{d_1, d_2^*\}, D(t_2) = \{d_3, d_4^*\}, D(t_3) = \{d_3, d_4^*\}, D(t_4) = \{d_1, d_3^*\}, \text{ and}$$

$$V_1 = \{ \langle s_1, t_1, \mathbf{0.2}, \mathbf{6} \rangle, \langle s_1, t_3, 0.2, 7 \rangle, \langle s_2, t_2, \mathbf{0.2}, \mathbf{4} \rangle, \langle s_2, t_1, 0.2, 5 \rangle, \langle s_3, t_1, \mathbf{0.35}, \mathbf{8} \rangle, \langle s_3, t_2, 0.35, 9 \rangle \}.$$

The top votes are marked above using bold fonts, and the top transactions are t_1 and t_2 (recall that only the top votes and transactions are considered in the commitment decision at any stage). Server s_1 commits t_1 , since $votes(t_1) = 0.55$. No candidate transactions become obsolete in this case, as t_1 does not conflict with any other candidate. Therefore, $V_1 = \{ \langle s_1, t_3, \mathbf{0.2}, \mathbf{7} \rangle, \langle s_2, t_2, \mathbf{0.2}, \mathbf{4} \rangle, \langle s_3, t_2, \mathbf{0.35}, \mathbf{9} \rangle \}$. Transaction t_2 is *still* a top transaction and t_3 has also become a top transaction. At this point, t_2 commits since $votes(t_2) = 0.55$. Server s_1 then aborts t_3 , which has become obsolete, and discards the corresponding votes. Finally $V_1 = \{ \}$.

after it is certified by all servers (similar to Agrawal's protocol [3]). This latter approach always requires contacting all servers in the system, which may be a serious restriction during times of low availability.

5. Support for mobility

This paper focuses on the performance of Deno’s commit protocols. For completeness, however, we briefly summarize some of Deno’s mobility-specific features.

Proxies

Deno allows servers to specify proxies to represent them during planned disconnections (during an airplane trip, for example) by voting in their place. The use of proxies can prevent degradation in the overall commit rate when devices have expected, planned-for disconnections. In fact, proxies can even improve commit latency because currency is concentrated in fewer servers, and fewer rounds of communication are required to establish a quorum. Proxies are implemented by having the proxy tell other servers that the primary’s vote is the same as it’s own while the proxy is engaged. A proxy vote is then indistinguishable to other servers from the situation where a server votes and then disconnects. When a primary reconnects, it updates its own information to match that of the proxy, including votes on prior and current tentative updates. The primary treats any votes cast in its behalf as if they had been cast directly. In particular, any votes cast for tentative updates remain cast. The result is that there are no race conditions, and the entire proxy engagement is transparent to the rest of the system.

Currency management

This paper assumes a uniform distribution of currency across all servers. In fact, the system initially gives all currency to the server that created the objects. Other servers obtain currency along with their initial copies of the data. Since the total amount of currency in the system always remains constant, a server that provides currency to another must lose currency of the same amount. In general, servers do not know the number of servers in the system, and so can not create an initial uniform distribution. However, the use of subsequent peer-to-peer currency exchanges allows the system to quickly approach any target distribution [13].

A second major question is the issue of how to deal with failures. Deno detects failures and network partitions through standard timeout mechanisms. Machines are declared “dead” by committing a retirement transaction that revokes the dead machine’s currency and redistributes it to other servers. This is effectively implemented as an involuntary proxy assignment.

Access control

The design of Deno’s security architecture includes authenticating servers using public-key cryptography when they first request copies of the database. All subsequent server-to-server communication is encrypted with private-key cryptography using a per-session key, where a session is defined to be the life of a particular database. However, this architecture is not fully implemented, and is not reflected in the performance data below.

Application-specific commutativity information

Disconnected and weakly-connected environments can be characterized by disconnections, high communication latencies, and incomplete connectivity. Therefore, applications running on top of these environments and systems need be designed so as to minimize conflicts among updates in order to avoid high abort rates [19]. One approach to accomplish this is to have the applications export domain-specific semantic information that can be used by the underlying system to modify application’s consistency requirements [34]. To this end, Deno’s extended protocol supports *commutativity procedures* to exploit application-specific commutativity information. A commutativity procedure is a simple query over the data-

base specifying an acceptance criterion [19]. If the query is satisfied, the transaction is considered as valid with respect to the current state of the database (event though it may already become obsolete in the traditional sense).

Deno executes a transaction's commutativity procedure (if it exists) if and when it becomes obsolete. If the acceptance criterion implemented by the procedure is satisfied, the transaction is kept alive and not aborted. Note that the use of commutativity procedures does not in any way change the consistency guarantees provided by the Deno's protocols. The protocols still ensure a consistent global ordering of transactions. This guarantee is the reason that commutativity procedures are only supported by the extended protocol. The base protocol does not guarantee a global ordering of all transaction commits, so the view of the database seen by the commutativity procedures, and the answer returned by any such procedure, would be site-dependent.

6. Performance evaluation

This section describes the performance of the Deno prototype. Note that the primary advantage gained in combining weighted voting with epidemic information flow is in increased availability. Rather than belabor the obvious, we instead investigate the performance impact of such a combination. Our earlier work [25, 26] presented simulation results showing the potential for good performance on single-object updates. This work presents results for multi-object transactional updates on a real system. Additionally, this work introduces and characterizes the performance impact of speculative voting and information dissemination.

6.1 Experimental methodology

We performed the experiments on a cluster of 15 Linux machines, each running a single copy of the Deno server. Each machine has two 400 MHz Pentium II's, and 256 MBytes of memory. However, none of the protocols discussed below ever came close to consuming all of a machine's resources. We have intentionally set our communication rates low in order to reflect the constraints of our expected environments. Instead, our performance evaluation concentrates on relative performance by comparing representative protocols.

The machines were connected via a 100Mbps Ethernet network and the Deno servers communicated using UDP packets. In order to concentrate on the convergence speed of the protocols, we used a small database consisting of 100 data objects of size 20K each. Each Deno server periodically initiates a synchronization session (with a given synchronization period of 5.0 secs) by sending a *pull* request to another randomly selected Deno server. Our experimental testbed differs from the real world in that the set of servers is constant, and assumed to be well-known. This distinction should not affect our findings on relative performance, but partial information about other servers can hurt performance.

Each server generated transactions according to a global transaction rate (specified relative to a synchronization period). Each transaction accessed and modified up to five data items. Since our focus is on the performance of the global consistency protocol, we did not model any read-only transactions. Currency is uniformly distributed across servers in all of the experiments. All objects are replicated at all servers. The main parameters and settings used in the experiments are summarized in Table 1.

The results presented in the following graphs are the average of five independent runs of executing 1000 transactions in the system. The contributions of the first 50 transactions are excluded to account to eliminate system *warm-up* effects.

Note that the bandwidth requirements for transactional and consistency data were negligible compared to that required for propagating updated values, so we do not consider this question further.

Parameter	Description	Setting
Synch Period	Mean synchronization period (uniform)	0 – 5 (secs)
Transaction Rate	Mean transaction generation rate (uniform)	0 – 25 (trans/synch period)
Num Servers	Number of Deno servers	3 – 15
Trans Size	Number of items updated by a transaction (uniform)	0 - 5
Commutativity Ratio	The probability that a transaction is acceptable on a given db state	0 – 1

Table 1: Primary experimental parameters and settings

6.2 Protocols evaluated

We now investigate the performance characteristics of our protocols using our prototype. We look at two versions of Deno’s protocol, `Deno-weak` (Section 3), and `Deno-strong` (Section 4). Additionally, we investigate two representative epidemic replication schemes from the literature. The first, `primary`, is an epidemic primary-copy scheme that uses a specialized primary server to serialize the updates, while propagating the updates using epidemic flow. This protocol is similar to that used in Bayou [34], but does not include advanced Bayou features such as *dependency checks* and *merge procedures*. In our implementation, we modeled this scheme simply by allocating all the currency at a single server. Note that primary-copy protocols trade availability for a presumed advantage in performance. One of our findings is that this performance advantage is not significant in protocols that use epidemic-style communication.

The second scheme, `write-all`, is a “Read-One, Write-All” (ROWA) [9] protocol, where servers can only commit transactions after ensuring that all other servers are ready to commit. Therefore, a transaction has to be propagated to all the servers before it can be committed. Furthermore, when a server receives conflicting transactions, it has to abort all of those transactions in order to ensure global consistency. A similar epidemic ROWA protocol was proposed by Agrawal *et al.* [3].

6.3 Commit delays

We first investigate the update propagation characteristics of different epidemic protocols. We begin by examining the *first commit delay*, which is the traditional commit delay denoting the time between the initiation of a transaction and the time it is *first* committed at a server. Figure 4(a) shows the first commit delays for the `primary`, `write-all`, and Deno. In `primary`, a transaction commits when the primary server receives the transaction (unless the primary decides to abort the transaction). Therefore, such a scheme clearly (*first*) *commits* a transaction much faster than a uniform voting approach, which requires a transaction to visit at least a majority of servers (assuming a uniform currency distribution). On the other hand `write-all` requires all servers to receive and certify the transaction, and performs poorly compared to either Deno protocol or `primary`.

Unlike traditional synchronous environments where transactions are committed synchronously at every server, commit times typically exhibit wide variability in asynchronous environments. The time at which the *first* server commits a transaction is, thus, not necessarily the quantity that best predicts application performance with epidemic information propagation. Since typically all servers have an equal chance of being accessed, an equally useful metric would be the *average commit time*, the time to commit a transaction averaged over all servers. Figure 4(b) presents the corresponding average commit delays for the `primary`, `write-all`, and Deno protocols. We see that `write-all` still performs significantly worse than the other protocols, and we also observe that the gap between Deno and `primary` does not exist anymore.

We further explore the reason behind the good performance of the Deno protocol by plotting the number of servers that committed the transaction as time progresses (for 15 servers) in Figure 5. Although

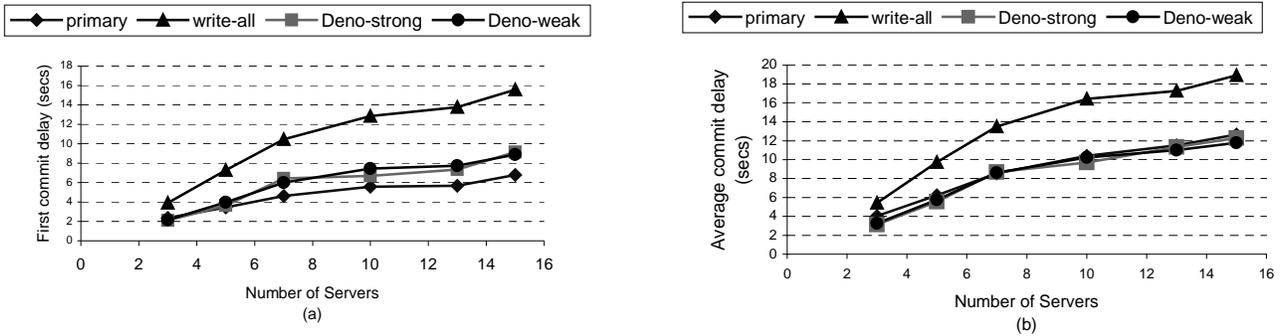


Figure 4: Transaction commit delays versus system size

the *primary* server commits the transaction quickly, this information propagates to other servers relatively slowly. This is because all other servers must learn of the commitment, directly or indirectly, from the primary server. With the Deno protocols, on the other hand, distinct servers may either learn the commitment from other servers (as in the case of *primary*), or commit the transaction *independently*. In the presented example, for instance, about seven servers (on the average) committed the transaction independently. The delay between the first and subsequent commits is thus quite small, as revealed by the high slope of the *Deno-weak* curve in Figure 5. In an environment where updates are propagated asynchronously, average commit delay is as important as the first commit delay because committed transactions are only useful at a server when reflected in the local replicas. One important implication is that the performance penalty of using voting rather than a primary-copy approach is not as large as commonly assumed in the kinds of environments we address.

However, the most interesting results from this series of experiments is that the basic Deno protocol did not appear to have any advantage of the extended version. We had expected the stronger semantics of the extended protocol to exact a performance cost. Instead, the difference between the commit delays of the two with little contention appears to be negligible (as shown in Figure 4), and is only an average of 10% with significant contention. The case with contention was where we expected the most degradation in performance, as the requirement of a global ordering effectively increases the number of conflicts. This increase in conflicts, in turn, forces more currency to be inspected before a winner of a given “election” can be determined. For example, we only need >50% of the currency in order to determine the winner of an election if there are no conflicting transactions, but we may need all of the currency in order to decide between two or more. However, the increase in required *currency* is offset by an increase in *concurrency*.

Consider the process of committing a transaction with no contention. Notice of the transaction has to propagate to half of the system servers before it can be committed. With two conflicting transactions that gain votes at the same rate, on the other hand, all of the votes may need to be cast before a winner is identified. However, each transaction can collect server votes independently, with notice of all votes arriving at some intermediate server in approximately the same amount of time as was needed for the single transaction case. Therefore, update contention does not necessarily increase commit delays.

6.4 Effects of contention

The previous subsection focused on the speed of transaction commits protocol performance when there is no update contention. This section studies the effects of transaction generation rate on the overall performance of the system.

Figure 6 presents the performance results of the protocols *under update contention*. More specifically, the figure shows the *commit percentage* (i.e., the percentage of initiated transactions that are committed)

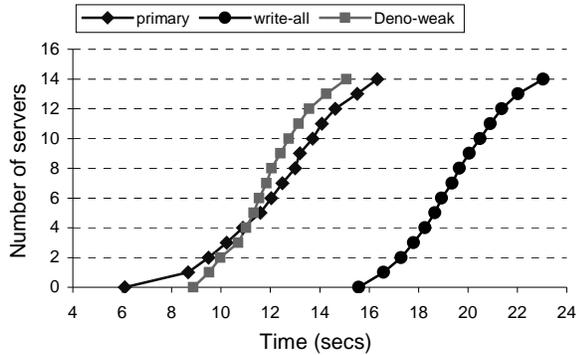


Figure 5: Number of servers that committed the transaction as time progresses (15 servers, Synch Period=5.0 secs)

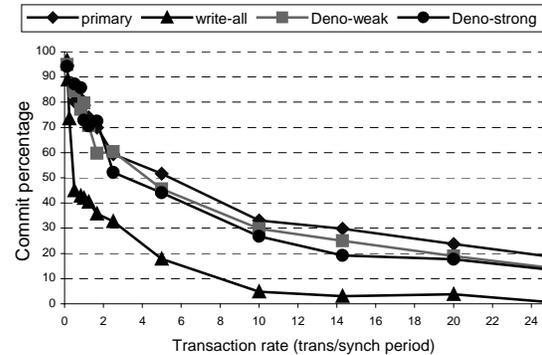


Figure 6: Commit percentages (15 servers, Synch Period=5.0 secs)

results for different levels of transaction generation rate (for 15 servers) for all protocols. The figure shows that all approaches suffer from the increased transaction rate due to the global update consistency requirement that only one out of a set of conflicting transactions can commit. Under very small transaction rates ([0.0-1.0] trans/synch period), all protocols perform fairly well, achieving commit percentages of around 100%. With increasing transaction rates, however, commit percentages drop for all protocols significantly. We observe the most dramatic fall for `write-all`: at around a transaction rate of 1.0, the commit percentage of `write-all` is less than 50%, whereas the commit percentages of the other protocols are above 70%. Overall, `primary` achieves the best commit percentage, followed closely by the weak and strong versions of Deno. The difference between the two versions of Deno as well as the difference between Deno protocols and `primary` over the whole range shown is small (within absolute 5%). The performance of `write-all` is significantly lower than the rest of the protocols. In fact, at (and beyond) a transaction rate of 25.0 (not shown), `write-all` does not commit any transactions. The main reason for this difference is that `write-all` has to abort all conflicting transactions, as it is not equipped with any mechanisms to globally single out a transaction to commit (out of a set of conflicting transactions). The other protocols continue to commit transactions *regardless* of the transaction rate (not shown).

6.5 Speculative voting and update propagation

Recall from Section 3 that a transaction that completes its execution is blocked until the local server has decided whether to commit or abort all conflicting candidate transactions. Blocked transactions can only proceed and participate in the global voting protocol after the conflicting transactions are terminated. Such a conservative blocking approach is used by many pessimistic distributed protocols [9, 20].

An optimistic alternative is for the blocking phase to be skipped by having servers immediately vote for all transactions as soon as they finish their local execution. These transactions are immediately candidates to be added to subsequent anti-entropy sessions. The advantage of such speculative voting is that transactions can make progress, in terms of gathering votes, *while* the system is still deciding the fate of prior transactions. Speculative votes are useful when previous conflicting transactions are aborted. As shown below, the advantage conferred by this technique is larger when there are commuting transactions in the system.

Example 5: In order to illustrate the potential benefits of speculation, consider a system of three servers, s_1 , s_2 , and s_3 , and two conflicting transactions, t_1 and t_2 . Assume that both transactions complete execution at s_1 in the order presented. In the (standard) non-speculative protocol, t_2 blocks until t_1 gets

propagated through the system, either gets committed or aborted, and this termination information is received (or made independently) at s_1 . If t_1 gets committed, then no sacrifice is made because t_2 is aborted in any case. On the other hand, if t_1 gets aborted, t_2 will resume and the whole process will be repeated to decide on the fate of t_2 . If speculative propagation is used, however, t_2 will be propagated along with t_1 and will gather votes that will be immediately available for use in case t_1 gets aborted. In this case, speculation will cut down the commit delays significantly by making progress during an interval otherwise wasted waiting for other transactions to terminate. The cost of speculation is that some transactions that will eventually get aborted will also be propagated through the system unnecessarily, resulting in a waste of communication bandwidth. We examine this tradeoff below.

In order to support speculative transaction propagation and voting, we extended our protocols such that servers vote for locally completed transactions immediately (without blocking after any other transaction) and propagate them through the system. In a sense, the system makes available the transactions and votes that may be required in future commitment processes speculatively, without knowing whether they will be necessary or valid.

In order to support speculation in Deno, we made the following modifications to the voting rule: Whenever a transaction locally completes execution, the server immediately votes for it as if it is a candidate transaction received from another server, inserting the transaction to the candidate list. More specifically, when a transaction t locally completes its execution, the server votes *yes* for t if there is not another conflicting candidate transaction. Otherwise, the server votes *no* for t . Notice that this change eliminates the need for a blocked transaction list, as all local transactions become candidates as soon as they locally complete execution.

Interestingly, we do not need to make any modifications to the commit rules we described earlier. The two commit rule versions, both the weak- and the strong-consistency versions, transparently decide on which votes should be considered in the current commitment decision, thereby ignoring the speculated votes for the current commit decision. The strong-consistency rule utilizes timestamp information (already available at the votes) to take into account the only the *top* votes as current votes. The weak-consistency rule implicitly utilizes *yes* and *no* votes. For each transaction t , only the votes from a given server that have lower timestamps than the timestamp of that server's vote for t are used in the current commit decision.

Figure 7 examines the benefits of speculative update propagation and voting for varying degrees of commutativity by showing the performance of speculative (Deno-spec) and non-speculative (Deno-nonspec) versions of Deno-strong. A commutativity ratio of .25, for example, implies that 25% of transactions made obsolete (in the sense discussed in Section 3) run commutativity procedures whose acceptance criteria are satisfied. Somewhat non-intuitively, larger commutativity ratios result in larger commit delays for the non-speculative version of the protocol. The reason is that increasing commutativity results in fewer aborted transactions, which in turn increases contention for those transactions that have not been aborted.

By contrast, Deno-spec's commit delay is largely constant across all commutativity ratios. Speculative voting confers a performance advantage of about 15% even with a commutativity ratio of 0.0 (i.e., the default case where no transactions commute). The gap increases with commutativity ratio until Deno-nonspec's commit delay is more than twice Deno-spec's at a ratio of 1.0.

As discussed above, the benefits of speculation come at the expense of propagating more transactions and votes. To this end, we also investigated the relative bandwidth requirements of both approaches. We

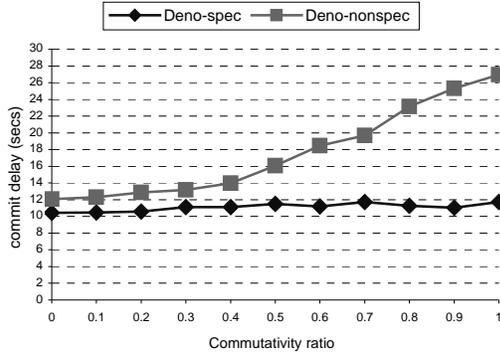


Figure 7: Speculation effects on commit delay (15 servers, Synch Period=5.0 secs)

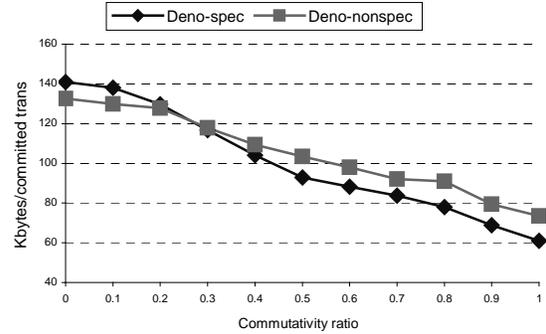


Figure 8: Speculation effects on bandwidth usage (15 servers, Synch Period=5.0 secs)

found out that `Deno-spec` indeed sends about 6% larger messages on the average (which turns out to be about 200 bytes/message more than the non-speculative case on average in our setting). Although the speculative approach uses more bandwidth per message, as demonstrated in the previous experiment, the commit rates for different approaches are quite different and need to be taken into account when comparing the relative bandwidth *utilizations* (e.g., how much useful work is accomplished per byte transmitted.) Figure 8 shows the amount of information sent across all servers (in KBytes) per committed transaction for `Deno-spec` and `Deno-nonspec`. For low commutativity ratios (up to .10), `Deno-spec` propagates about 4-6% more information per transaction commit. At a commutativity ratio of .20, both approaches propagate almost the same amount of information. Beyond this ratio, the speculative protocol sends less information than the non-speculative version, with the difference increasing as the commutativity increases. At a commutativity ratio of 1.0, `Deno-spec` propagates about 16% less information per committed transaction.

To summarize, the speculative version not only decreases average commit delays, but it also decreases bandwidth requirements per committed transaction.

7. Related work

The problem of consistent access to replicated data has long been studied in many contexts and a wide variety of solutions have been proposed, e.g., [1, 2, 9, 14, 17, 32, 35] (see [20] for a compact survey of replication techniques in distributed environments). Due to the intrinsic shortcomings of traditional synchronous replication solutions [18, 32, 35] in large-scale, mobile, and weakly-connected environments [19], asynchronous replication protocols have recently gained a lot of attention (e.g., [5, 11, 12, 24, 27]). Asynchronous approaches commonly allow servers to execute updates locally without any synchronization with other servers, and propagating updates afterwards as separate activities.

According to Gray *et al.* [19], existing asynchronous solutions fall into two broad categories, *master* and *group*, with respect to how they regulate update creation and data item ownership. Master approaches require a data item update to be first performed at the *primary copy* of the item. The updates are then propagated to *secondary copies*. For instance, Breitbart *et al.* [11] recently proposed a master asynchronous approach that ensure serializable executions, and demonstrated significant performance improvements over master synchronous protocols in many cases. Group approaches, also called “update-anywhere”, allow any copy to be updated, thereby yielding maximum flexibility with respect to where and when to create updates. This flexibility is highly desirable in many collaborative groupware applications (e.g., CAD, CASE, Lotus Notes [24], etc.), especially when operating under mobile and weakly-

connected environments [19]. The cost of this flexibility is, however, increased update conflicts, and the complexity and overhead of maintaining consistency.

Many existing asynchronous “update-anywhere” protocols typically utilize the epidemic model [3, 15, 24, 29, 31, 34]. Many epidemic systems take an optimistic approach and use reconciliation-based protocols (e.g., Ficus [29], Lotus Notes [24]) that are only viable in non-transactional single-item domains such as file systems. These approaches only ensure that all copies of a single item eventually converge to the same value, and therefore are not safe for environments requiring transactional semantics. For example, Rabinovich *et al.* [31] discussed an epidemic approach that enables eventual consistency of the replicas in a non-transactional framework. The main focus of the work was, however, on minimizing the overhead of the periodic synchronization sessions.

Bayou [34] takes a more pessimistic approach and ensures that all committed updates are serialized in the same order at all servers using a *primary-copy* scheme. More recently, Agrawal *et al.* [3] proposed a pessimistic “Read-One, Write-All” (ROWA) [9] approach that was the first epidemic protocol to ensure strong consistency and serializability. This protocol allows a transaction to commit only after all servers “certify” the transaction. Our protocols differ from these protocols primarily in using a novel combination of weighted-voting and epidemic information flow [15] to improve availability and performance. In Section 3.2, we also compare the performance of our protocol with variants of both primary-copy and ROWA type epidemic approaches.

The impact of mobility on data management and replication has been discussed in [4, 6, 7, 16, 22]. Pitoura and Bhargava proposed a two-level, cluster-based consistency model for intermittently-connected environments [30]. Barbara and Imielinski investigated different cache invalidation strategies for mobile wireless environments [8]. Huang *et al.* analyzed different data replication methods with the objective of minimizing the communication cost between the mobile and stationary hosts in a mobile environment [21].

Finally, we note that recent work [36] investigated why voting systems (i.e., quorums) have yet to become widespread in real-world applications. One of the conclusions is that voting does not enhance availability because either failures are positively correlated (when servers are on a single LAN) or network partitions occur (when servers are distributed across multiple LANs). In the latter case, a quorum constructed on a single LAN has higher availability than quorums constructed across multiple LANs. However, the weakly-connected environments addressed in this work fit neither category. Most failures (e.g., disconnections) are likely to be independent, and network partitions, while possible, are not the dominant cause of unavailability.

8. Conclusions

We have presented the design, implementation, and performance of Deno, a highly-available object-replication system that provides transactional semantics in mobile and weakly-connected environments. Deno’s consistency protocols are based on an asynchronous weighted-voting approach implemented through epidemic information flow. Our voting approach has significant advantages over previous schemes in that it achieves higher availability than primary-copy approaches [34], and higher availability and performance than ROWA approaches [3].

Our base protocol ensures weakly-consistent executions [9, 10, 17], where update transactions are serializable and queries always access transactionally-consistent database states. Our extended protocol provides strong consistency [9, 10, 17] and globally serializable executions by providing a unique global commit order on all update transactions. Both protocols allow queries to be executed and committed en-

tirely *locally*, and *without blocking*. Furthermore, neither protocol suffers from local or global deadlocks (see the Appendix for details).

To a large degree, our focus on mobility and weakly-connected environments dictated both the policies that we enforce and the mechanisms that we use to pursue policy goals. To summarize, our focus on mobility led directly to the following decisions:

- i) availability *paramount* → weighted voting
- ii) dynamic network topology and connectivity → epidemic information flow
- iii) new applications + weak connectivity → performance/semantics tradeoff, commut. procs.

We performed a detailed study of the Deno prototype's performance, particularly in regards to commit performance versus other protocols designed for similar arenas. One of the focuses of this study is the performance impact of the above tradeoffs. Comparing uniform voting and primary-copy variants allows us to characterize the cost of the higher availability provided by voting. Comparing the performance of our base and extended protocols allows us to assess the cost of the stronger transactional semantics.

Our main results are the following. First, the presumed performance advantage of the primary-copy approach over a uniform voting approach is nearly non-existent with asynchronous epidemic protocols. Average commit delays and commit percentages are nearly identical for the primary-copy protocol and even the basic, non-speculative version of the Deno protocol. The reason is that epidemic voting protocols allow servers to independently arrive at the same conclusions, whereas primary-copy schemes require all commit information to emanate from a single, distinguished server.

Second, our extended protocol performs nearly as well as the base protocol, while providing significantly stronger semantics. The lack of a performance advantage for a protocol with weaker semantics and fewer restrictions on update commits is perhaps our most surprising finding. The result is increased functionality at little cost in performance.

Finally, speculative update propagation and voting provides a considerable performance advantage for protocols that use pair-wise communication, and this advantage is magnified when application-specific commutativity information is used to decrease the rate of transaction aborts. This latter point is especially important as the database community is coming to the realization that non-transparent replication is crucial to good performance in dynamic and mobile environments [33].

This paper has assumed a flat organization of Deno servers. As with any distributed system, flat organizations, especially peer-to-peer models, suffer from low scalability. Significant performance improvements can be attained when a hierarchical synchronization topology is used to organize servers. Support for hierarchical organizations will form the basis of our future work.

References

- [1] A. E. Abbadi and S. Toueg, "Availability in Partitioned Replicated Databases," in *Proc. of the fifth ACM SIGACT-SIGMOD symposium on Principles of database systems*, 1986.
- [2] D. Agrawal and A. E. Abbadi, "An Efficient and Fault-Tolerant Solution for Distributed Mutual Exclusion," *ACM Transactions on Computing Systems*, vol. 9, pp. 1-20, 1991.
- [3] D. Agrawal, A. E. Abbadi, and R. Steinke, "Epidemic Algorithms in Replicated Databases," in *Proceedings of the Sixteenth ACM SIGACT-SIGMOD-SIGART Symposium on Principles of Database Systems*, May 1997.
- [4] R. Alonso and H. F. Korth, "Database System Issues in Nomadic Computing," in *Proc. of the ACM SIGMOD Int. Conf. on Management of Data*, Washington, DC, May 1993.
- [5] T. Anderson, Y. Breitbart, H. F. Korth, and A. Wool, "Replication, Consistency, and Practicality: Are These Mutually Exclusive?," in *In Proceedings of ACM SIGMOD International Conference on Management of Data*, 1998.
- [6] D. Barbara, "Mobile Computing and Databases," *TKDE*, vol. 1, pp. 108-117, 1999.
- [7] D. Barbara and H. Garcia-Molina, "Replicated Data Management in Mobile Environments: Anything New Under the Sun?" in *IFIP Working Conference on Applications in Parallel and Distributed Computing*, April 1994.
- [8] D. Barbara and T. Imielinski, "Sleepers and Workaholics: Caching strategies in mobile environments," in *Proc. of the ACM SIGMOD Int. Conf. on Management of Data*, 1994.
- [9] P. A. Bernstein, V. Hadzilacos, and N. Goodman, *Concurrency Control and Recovery in Database Systems*. Reading, Massachusetts: Addison-Wesley, 1987.
- [10] P. Bober and M. Carey, "Multiversion Query Locking," in *Proc. of the VLDB Conference*, British Columbia, Canada, 1992.
- [11] Y. Breitbart, R. Komondoor, R. Rastogi, S. Seshadri, and A. Silbershatz, "Update Propagation Protocols for Replicated Databases," in *Proc. of the ACM SIGMOD Int. Conf. on Management of Data*, Philadelphia, PA, 1999.
- [12] Y. Breitbart and H. F. Korth, "Replication and Consistency: Being Lazy Helps Sometimes," in *Proc. of the Symposium on Principles of Database Systems*, Tucson, Arizona, May 1997.
- [13] U. Cetintemel and P. J. Keleher, "Light-Weight Currency Management Mechanisms in Deno," in *The 10th IEEE Workshop on Research Issues in Data Engineering (RIDE2000)*, February 2000.
- [14] S. Davidson, H. Garcia-Molina, and D. Skeen, "Consistency in a Partitioned Network: A Survey," *ACM Computing Surveys*, vol. 17, pp. 341-370, 1985.
- [15] A. Demers, D. Greene, C. Hauser, W. Irish, J. Larson, S. Shenker, H. Sturgis, D. Swinehart, and D. Terry, "Epidemic algorithms for replicated database maintenance," in *Proc. of the Symposium on Principles of Distributed Computing*, 1987.
- [16] M. Dunham and A. Helal, "Mobile Computing and Databases: Anything New?," *SIGMOD Record*, vol. 24, pp. 5-9, 1995.
- [17] H. Garcia-Molina and G. Wiederhold, "Read-Only Transactions in a Distributed Database System," *ACM Transactions on Database Systems*, vol. 7, pp. 209-234, June 1982.
- [18] D. K. Gifford, "Weighted Voting for Replicated Data," in *Proc. of the ACM Symposium on Operating Systems Principles*, 1979.

- [19] J. Gray, P. Helland, P. O’Neil, and D. Shasha, “The Dangers of Replication and a Solution,” in *Proceedings of the 1996 ACM SIGMOD international conference on Management of data*, June 1996.
- [20] A. A. Helal, A. A. Heddaya, and B. B. Bhargava, *Replication Techniques in Distributed Systems*: Kluwer Academic Publishers, 1996.
- [21] Y. Huang, A. P. Sistla, and O. Wolfson, “Data replication for mobile computers,” in *Proc. of the ACM SIGMOD Int. Conf. on Management of Data*, 1994.
- [22] T. Imielinski and B. R. Badrinath, “Wireless Mobile Computing: Challenges in Data Management,” *Communications of the ACM*, vol. 37, pp. 19-28, October 1994.
- [23] S. Jajodia and D. Mutchler, “Dynamic Voting Algorithms for Maintaining the Consistency of a Replicated Database,” *ACM Transactions on Database Systems*, vol. 15, pp. 230-280, 1990.
- [24] L. Kawell, S. Beckhardt, T. Halvorsen, R. Ozie, and L. Greif, “Replicated Document Management in a Group Communication System,” in *Proc. of the Conf. on Computer Supported Cooperative Work*, 1988.
- [25] P. Keleher and U. Cetintemel, “Consistency Management in Deno,” *The Journal on Special Topics in Mobile Networking and Applications (MONET)*.
- [26] P. J. Keleher, “Decentralized Replicated-Object Protocols,” in *The 18th Annual Symposium on Principles of Distributed Computing (PODC ‘99)*, May 1999.
- [27] R. Ladin, B. Liskov, L. Shrira, and S. Ghemawat, “Providing High Availability Using Lazy Replication,” *ACM Transactions on Computing Systems*, vol. 10, pp. 360-391, November 1992.
- [28] F. Mattern, “Virtual Time and Global States of Distributed Systems,” in *Parallel and Distributed Algorithms*, Amsterdam, 1989.
- [29] T. W. Page, R. G. Guy, J. S. Heidemann, D. Ratner, P. Reiher, A. Goel, G. H. Kuenning, and G. J. Poppek, “Perspectives on Optimistically Replicated Peer-to-Peer Filing,” *Software—Practice and Experience*, vol. 28, pp. 155-180, February 1998.
- [30] E. Pitoura and B. Bhargava, “Maintaining Consistency of Data in Mobile Distributed Environments,” in *Proc. of the International Conference on Distributed Computing Systems*, May 1995.
- [31] M. Rabinovich, N. H. Gehani, and A. Kononov, “Scalable Update Propagation in Epidemic Replicated Databases,” in *International Conference on Extending Database Technology (EDBT)*, 1996.
- [32] M. Stonebraker, “Concurrency control and consistency of multiple copies of data in distributed INGRESS,” *IEEE Transactions on Software Engineering*, vol. SE-5, pp. 188-194, May 1979.
- [33] D. B. Terry, K. Petersen, M. J. Spreitzer, and M. M. Theimer, “The Case for Non-Transparent Replication: Examples from Bayou,” in *IEEE Data Engineering*, December 1998.
- [34] D. B. Terry, M. M. Theimer, K. Petersen, A. J. Demers, M. J. Spreitzer, and C. H. Hauser, “Managing Update Conflicts in a Weakly Connected Replicated Storage System,” in *Proc. of the ACM Symposium on Operating Systems Principles*, 1995.
- [35] R. H. Thomas, “A Majority Consensus Approach to Concurrency Control for Multiple Copy Databases,” *ACM Transactions on Database Systems*, vol. 4, pp. 180-209, 1979.
- [36] A. Wool, “Quorum Systems in Replicated Databases: Science or Fiction?,” *Bulletin of the Technical Committee on Data Engineering*, vol. 21, pp. 3-11, 1998.

Appendix

Termination Issues

Lemma 4 (*Eventual Termination – Base Protocol*) A candidate transaction eventually terminates (i.e., commits or aborts).

Proof (*Sketch*): Suppose there exists a candidate transaction t that never terminates. We can partition the set of servers into three disjoint subsets as (1) the servers that voted *yes* for t , $yes(t)$, (2) servers that voted *no* for t , $no(t)$, and (3) servers that have not yet observed t , denoted $unknown_servers(t)$. Assuming that information eventually propagates to all servers, $unknown_servers(t)$ will eventually become empty. Let the conflict set of t , $CS(t)$, denote the set of candidate transactions that conflict with t . When $unknown_servers(t)$ becomes empty, $CS(t)$ can not grow further due to the voting rule (see Section 3.1.2), since all servers voted for either t or another transaction that conflicts with t . Now consider the case where all candidate transactions $t' \in CS(t)$ are observed at all servers. At this point, $votes(t)$ and $votes(t')$ for all $t' \in CS(t)$ are determined. As a result, $unknown(t)$ and $unknown(t')$ for all $t' \in CS(t)$ are all 0.0. Therefore, the commit rule will commit the transaction with the most votes (or in the case of a tie the one executed at the server with the smallest id) and the rest will be aborted, thereby contradicting our initial claim. Moreover, a deadlock situation due to vote dependencies can not exist. Such a deadlock has to involve a cycle of the form $votes(t_1) > votes(t_2) > \dots > votes(t_n) > votes(t_1)$ where t_1, t_2, \dots, t_n are candidate transactions. Since both $votes(t_1) < votes(t_n)$ and $votes(t_n) < votes(t_1)$ can not be true at the same time, we conclude that such a deadlock cannot exist.

Now consider a blocked transaction t . Transaction t will eventually become a candidate since 1) the set of candidate transactions that t is blocked after will all eventually terminate (see earlier discussion), and 2) the blocked transactions are considered in the order they are entered into the blocked list, so t is not going to wait indefinitely before being considered for candidacy.

Lemma 5 (*Eventual Termination – Extended Protocol*) A candidate transaction eventually terminates (i.e., commits or aborts).

Proof (*Sketch*): We consider two cases. First consider a top transaction, t . Assume eventually that the top votes cast by all servers are known (i.e., $unknown=0.0$). Let U denote the non-empty set of top transactions that obtained more votes than the remaining top transactions. If U contains a single transaction, then that transaction commits. Otherwise, U contains a set of transactions and the top transaction whose execution server has the smallest id commits. If t is the transaction that commits, then we are done. Otherwise; if t becomes obsolete, then it gets aborted, or else t remains a top transaction. The process repeats and t either becomes obsolete and is aborted, or eventually gets enough top votes and commits. Now consider a *non-top* transaction, t . Transaction t will either become obsolete and get aborted by the commit of a top transaction, or eventually become a top transaction itself and terminate (by the discussion in the first case). Therefore, we conclude that a candidate transaction always terminates.

Correctness of the Extended Protocol (Section 4)

Lemma 6 (*Global Update Consistency*) The protocol presented above ensures a unique global commit order on the set of update transactions.

Proof (*Sketch*): In particular, we show that each server commits the same update transactions in the same order. Assume that t_i is the very first transaction that committed at server s . Extending the discussion presented in the proof of Lemma 1 by treating the top transactions to be the only conflicting transactions in

the system, we can conclude that t_i is the first transaction to commit at all servers. A straightforward induction on the sequence of committed transactions concludes the proof.

Theorem 2 (*Strong Consistency*) The protocol presented above provides strong consistency and serializability.

Proof (*Sketch*): Lemma 1 ensures that there are no update transaction cycles. Without loss of generality, assume that there is a multiple-query cycle of the form

$$q_1 \rightarrow t_1 \rightarrow q_2 \rightarrow t_2 \rightarrow \dots \rightarrow q_n \rightarrow t_n \rightarrow q_1.$$

Consider $q_1 \rightarrow t_1$, which implies that there is an item d read by q_1 and then updated by t_1 . By Lemma 3, q_1 commits before t_1 at the execution site of q_1 , say s_1 . Now consider $t_1 \rightarrow q_2$ and $q_2 \rightarrow t_2$, which together imply that t_1 commits before q_2 and, therefore, before t_2 at the execution site of q_2 , say s_2 . Therefore, by Lemma 1, t_1 commits before t_2 at all sites. Using a straightforward induction, we can say that t_1 commits before t_n at all sites. However, $t_n \rightarrow q_1$ implies that t_n commits before q_1 at s_1 , creating the contradiction that concludes the proof.