18. Consensus

Consensus (sometimes called ‘reliable broadcast’ or ‘atomic broadcast’) is a fundamental building block for distributed systems. Informally, we say that several processes achieve consensus if they all agree on some value. Three obvious applications are:

- Distributed transactions, where all the processes need to agree on whether a transaction commits or aborts. Each transaction needs a new consensus on its outcome.
- Membership, where a set of processes cooperating to provide a highly available service need to agree on which processes are currently functioning as members of the set. Every time a process fails or starts working again there must be a new consensus.
- Electing a leader of a group of processes.

Replicated state machines

There is a much more general way to use consensus, as the mechanism for coding a highly available state machine, which is the basic tool for building a highly available system. The way to get availability is to have either perfect components or redundancy. Perfect components are too hard, which leaves redundancy. The simplest form of redundancy is replication: have several copies or replicas of each component, and make sure that all the non-faulty components do the same thing. Since any computation can be expressed as a state machine, a replicated state machine can make any computation highly available.

Recall the basic idea of a replicated state machine:

If the transition relation is deterministic (in other words, is a function from (state, input) to (new state, output)), then several copies of the state machine that start in the same state and see the same sequence of inputs will do the same thing, that is, end up in the same state and produce the same outputs.

So if several processes are implementing the same state machine and achieve consensus on the values and order of the inputs, they will do the same thing. In this way it’s possible to replicate an arbitrary computation and thus make it highly available. Of course we can make the order a part of the value of the input by defining some total order on the set of possible inputs. We have already seen one application of this replicated state machine idea, in the code for transactions; there the replication takes the form of redoing a sequence of actions that is remembered in a log.

Suppose, for example, that we want to build a highly available file system. The transitions are read and write operations on the files (and rename, list, … as well). We make several copies of the file system and make sure that they process read and write operations in the same order. A client sends its operation to some copy, which gets consensus that it is the next operation. Then all the copies do the operation, and one of them returns the result to the client. You might think that a read could be handled by any copy with no need for consensus, since it doesn’t change the state of the file system. Without consensus, however, a read might fail to see the result of a write that finished before the read started, since the read might be handled by a copy whose state is behind the current state of the file system.

In many applications the inputs are requests from clients to the replicated service. Typically different clients generate their requests independently, so it’s necessary to agree not only on what the requests are, but also on the order in which to serve them. The simplest way to do this is to number them with consecutive integers, starting at 1. This is what is done in ‘primary copy’ replication, since it’s easy for one place (the primary) to assign consecutive numbers.

The literature is full of other schemes for achieving consensus on the order of requests when their total order is not derived from consecutive integers. These schemes label each input with some label from a totally ordered set (for instance, (client UID, timestamp) pairs) and then devise some way to be certain that you have seen all the inputs that can ever exist with labels smaller than a given value. They are complicated, and of doubtful utility.2 People who do it for money use primary copy.3

Unfortunately, consensus is expensive. The section on optimizations at the end of this handout explains a variety of ways to make a replicated state machine run efficiently: leases, transactions, and batching.

Spec for consensus

Here is the spec for consensus; we have seen it already in handout 8 on history and prophecy variables. The idea is that the outcome of consensus should be one and only one of the allowed values. In the spec there is an outcome variable initialized to nil, and an action Allow(v) that can be invoked any number of times. There is also an action Outcome to read the outcome variable; it must return either nil or a v which was the argument of some Allow action, and if it doesn’t return nil it must always return the same v.

More precisely, we have two requirements:

Agreement: Every non-nil result of outcome is the same.

Validity: A non-nil outcome equals some allowed value.

Validity means that the outcome can’t be any arbitrary value, but must be a value that was allowed. Consensus is reached by choosing some allowed value and assigning it to outcome. This spec makes the choice on the fly as the allowed values arrive.

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1 This approach was first proposed in a classic paper by Leslie Lamport: Time, clocks, and the ordering of events in a distributed system, Comm. ACM 21, 7, July 1978, pp 558-565. This paper is better known for its analysis of the partial ordering of events in a distributed system, which is too bad.


3 Jim Gray said this about using locks for concurrency control; see handout 20.
**MODULE Consensus** [V] EXPORT Allow, Outcome = % data value to agree on

VAR outcome : (V + Null) := nil

APROC Allow(v) = << outcome = nil >> outcome := v [] SKIP >>
APROC Outcome() -> (V + Null) = << RET outcome [] RET nil >>

END Consensus

Note that Outcome is allowed to return nil even after the choice has been made. This reflects the fact that in code with several replicas, Outcome is often coded by talking to just one of the replicas, and that replica may not yet have learned about the choice.

If only one Allow action occurs, there’s no need to choose a v, and the code’s only problem is to ensure termination. An algorithm that does so is said to implement ‘reliable’ or ‘atomic’ broadcast; there is only one sender, and either everyone or no one gets the message. The single Allow might not set outcome, which corresponds to failure of the sender of the broadcast message; in this case no one gets the message.

Here is an equivalent spec, slightly more complicated but perhaps more intuitive, and certainly closer to an implementation. It accumulates the allowed values and then chooses one of them in the internal action Agree.

**MODULE LateConsensus** [V] EXPORT Allow, Outcome =

VAR outcome : (V + Null) := nil
allowed : SET V := {}

APROC Allow(v) = << allowed / := {v} >>
APROC Outcome() -> (V + Null) = << RET outcome [] RET nil >>
% Only outcome is visible
APROC Agree() = << VAR v :IN allowed | outcome = nil >> outcome := v >>

END LateConsensus

It should be fairly clear that LateConsensus implements Consensus. An abstraction function to prove this, however, requires a prophecy variable, because Consensus decides on the outcome (in the Allow action) before LateConsensus does (in the Agree action). We saw these specs in handout 8 on generalized abstraction functions, where prophecy variables are explained.

In the code we have in mind, there are some processes, each with its own outcome variable initialized to nil. The outcome Variables are supposed to reach consensus, that is, become equal to the argument of some Allow action. An Outcome can be directed to any process, which returns the value of its outcome variable. The tricky part is to ensure that two non-nil outcome variables are always equal, so that the agreement property is satisfied.

We would also like to have the property that eventually Outcome stops returning nil. In the code, this happens when every process’ outcome variable is non-nil. However, this could take a long time if some process is very slow (or down for a long time).

We can change Consensus to express this with an internal action Done:

**MODULE TermConsensus** [V] EXPORT Allow, Outcome =

VAR outcome : (V + Null) := nil
done : Bool := false

APROC Allow(v) = << outcome = nil >> outcome := v [] SKIP >>
APROC Outcome() -> (V + Null) = << RET outcome [] done >>

THREAD Done() = << outcome # nil >> done := true >>

END TermConsensus

An even stronger spec returns an outcome only when it’s done:

APROC Outcome() -> (V + Null) = << done >> RET outcome [] RET nil >>

This is usually too strong for distributed code. It means that a process may not be able to respond to an Outcome request, since it can’t return a value if it doesn’t know the outcome yet, and it can’t return nil if anyone else has already returned a value.

**Facts about consensus**

In this section we summarize the most important facts about when consensus is possible and what it costs. You can learn more about this in Nancy Lynch’s course on distributed algorithms, 6.852J, or in her book cited in handout 2.

**Fault models**

To devise code for Consensus we need a precise model for the general setting of processes connected by links that can communicate messages from one process to another. In particular, the model must define what faults are possible. There are lots of ways to do this, and we have space to describe only the models that are most popular and closest to reality.

There are two broad classes of models:

- **Synchronous**, in which a non-faulty component makes its state transitions within a known amount of time. Usually this is coded by using a timeout, and declaring a component faulty if it fails to perform within the specified time.

- **Asynchronous**, in which nothing is known about the relative rate of progress of different components. In particular, a process can take an arbitrary amount of time to make a transition, and a link can take an arbitrary amount of time to deliver a message.

In general a process can send a message only to certain other processes; this “can send message” relation defines a graph whose edges are the links. The graph may be directed (it’s possible that A can talk to B but B can’t talk to A), but we will assume that communication is full-duplex so that the graph is undirected. Links are either working or faulty; a faulty link delivers no messages. Even a working link may lose messages, and in some models may lose any number of messages; it’s helpful to think of such a system as one with totally asynchronous communication.

Processes are either working or faulty. There are two models for a faulty process:
• **Stopping** faults: a faulty process stops making transitions and doesn’t start again. In an asynchronous model there’s no way for another process to distinguish a stopped process or link from one that is simply very slow.

• **Byzantine** faults: a faulty process makes arbitrary transitions; these are named after the Byzantine Empire, famous for treachery. The motivation for this model is usually not fear of treachery, but ignorance of the ways in which a process might fail. Clearly Byzantine failure is an upper bound on how bad things can be.

Is consensus possible (will it terminate)?

A consensus algorithm terminates when the outcome variables of all non-faulty processes equal some allowed value. Here are the basic facts about consensus in some of these models.

• There is no consensus algorithm that is guaranteed to terminate in an asynchronous system with perfect links and even one process that has a stopping fault. This startling result is due to Fischer, Lynch, and Paterson. It holds even if the communication system provides reliable broadcast that delivers each message either to all the non-faulty processes or to none of them. Real systems get around it by using timeout to make the system synchronous.

• Even in a synchronous system with perfect processes there is no consensus algorithm that is guaranteed to terminate if an unbounded number of messages can be lost (that is, if communication is effectively asynchronous). The reason is that the last message sent must be pointless, since it might be lost. So it can be dropped to get a shorter algorithm. Repeat this argument to drop all the messages. But clearly an algorithm with no messages can’t achieve consensus. The simplest case of this problem, with just two processes, is called the “two generals problem”.

• In a system with both synchronous processes and synchronous communication, terminating consensus is possible. If \( f \) faults are allowed, then:

   For processes with stopping faults, consensus requires \( f+1 \) processes and an \( f+1 \)-connected network (that is, at least one good process and a connected subnet of good processes after all the allowed faults have happened). Even if the network is fully connected, it takes \( f+1 \) rounds to reach consensus in the worst case.

   For processors with Byzantine faults, consensus requires \( 3f+1 \) processes, a \( 2f+1 \)-connected network, at least \( f+1 \) rounds of communication, and \( 2^f \) bits of data communicated.

   For processors with Byzantine faults and digital signatures (so that a process can present unforgeable evidence that another process sent it a message), consensus requires \( f+1 \) processes. Even if the network is fully connected, it takes \( f+1 \) rounds to reach consensus in the worst case.

The amount of communication required depends on the number of faults, the complexity of the algorithm, etc. Randomized algorithms can achieve better results with arbitrarily high probability.

Warning: In many applications the model of no more than \( f \) faults may not be realistic if the system is allowed to do the wrong thing when the number of faults exceeds \( f \). It’s often more important to do either the right thing or nothing.

**The simplest consensus algorithms**

There are two simple and popular algorithms for consensus. Both have the problem that they are not very fault-tolerant.

• A fixed ‘leader’, ‘master’, or ‘coordinator’ process that works like the consensus spec: it gets all the allow actions, chooses the outcome, and tells everyone. If it fails, you are out of luck. The abstraction function is just the identity on the leader’s state. TermConsensus.done is true iff everyone has gotten the outcome (or failed permanently). Standard two-phase commit for distributed transactions works this way.

• Simple majority voting. The abstraction function for outcome is the value that has a majority, or nil if there isn’t one. This fails if you don’t get a majority, or if enough members of a majority fail that it isn’t a majority any more. In the latter case you can’t determine the outcome. Example: \( a \) votes for 11, \( b \) and \( c \) vote for 12, and \( b \) fails. Now all you can see is one vote for 11 and one for 12, so you can’t tell that 12 had a majority.

**The Paxos algorithm: The idea**

In the rest of this handout, we describe Lamport’s Paxos algorithm for coding asynchronous consensus; Liskov and Oki independently invented this algorithm as part of a replicated data storage system. Its heart is the best asynchronous algorithm known, which is run by a set of leader processes that guide a set of agent processes to achieve consensus, correct no matter how many simultaneous leaders there are and no matter how often leader or agent processes fail and recover or how slow they are, and guaranteed to terminate if there is a single leader for a long enough time during which each member of a majority of the agent processes is up for long enough, but possibly non-terminating if there are always too many leaders (fortunate, since we know that guaranteed termination is impossible).

---


5 A graph is connected if there is a path (perhaps traversing several links) between any two nodes, and disconnected otherwise. It is \( k \)-connected if \( k \) is the smallest number such that removing \( k \) links can leave it disconnected.

To get a complete consensus algorithm we combine this with a sloppy timeout-based algorithm for choosing a single leader. If the sloppy algorithm leaves us with no leader or more than one leader for a time, the consensus algorithm may not terminate during that time. But if the sloppy algorithm ever produces a single leader for long enough the algorithm will terminate, no matter how messy things were earlier.

Paxos is the way to do consensus if you want a high degree of fault-tolerance, don’t have any real-time requirements, and can’t tightly control the time to transmit and process a message. There isn’t any simpler algorithm that has the same fault-tolerance. There is lots of code for consensus that doesn’t work.

The grand plan of the algorithm is to have a sequence of rounds, each with a single leader. This attacks the problem with simple majority voting, which is a single attempt to get a majority may fail victim to failure. Each Paxos round is a distinct attempt to get a majority. Each agent has a state variable \( s(a) \) that is a function of the round; that is, there’s a state value \( s(a)(n) \) for each round \( n \). To reduce clutter, we write this \( s_n^a \). In each round the leader:

- chooses the agents to learn their state for past rounds,
- commands the agents, trying to get a majority to accept \( v \), and
- if successful, distributes \( v \) as the outcome to everyone.

The outcome is the value accepted by a majority in some round. The tricky part of the algorithm is to ensure that there is only one such value, even though there may be lots of rounds.

Most descriptions of Paxos call the agents ‘voters’. This is unfortunate, because the agents do not make any decisions; they do whatever a leader requests, unless they have already done something inconsistent with that. In fact, an agent can be coded by a memory that has a compare-and-swap operation, as we shall see later. Of course, the leaders and agents can run on the same machine, and even in the same process. This is usually the way it’s coded, but the algorithm with separate leaders and agents is easier to explain.

It takes a total of \( 21/2 \) round trips for a successful round. If there’s only one leader that doesn’t fail, Paxos reaches consensus in one round. If the leader fails repeatedly, or several leaders fight it out, it may take arbitrarily many rounds to reach consensus.

The rounds are numbered (not necessarily consecutively) with numbers of type \( n \), and the numbers determine a total ordering on the rounds. Each round has a single value, which starts out \( \text{nil} \) and then may change to one of the allowed values; we write \( v_n \) for the value of round \( n \). In each round an agent starts out \( \text{neutral} \), and it can only change to \( v_n \) or \( \text{no} \). A \( v_n \) or \( \text{no} \) state can’t change. Note that different rounds can have different values. A round is \( \text{dead} \) if a majority has state \( \text{no} \), and \( \text{successful} \) if a majority has state \( v_n \). If a round is successful, that round’s value is the outcome.

The state of Paxos that contributes to the abstraction function to \text{LateConsensus} is

\[
\begin{align*}
\text{MODULE Paxos}[
\quad V, \\
\quad L \text{ WITH } \{\text{<}=: (L, L) \rightarrow \text{Bool}, \\
\quad A \text{ WITH } \{\text{majority} : \text{SET} A \rightarrow \text{Bool} \}
\}
\end{align*}
\]

\[
\text{TYPE I} = \text{Int}
\]

\[
\begin{align*}
N &= \{i, l\} \text{ WITH } \{\text{<}=: \text{LEQn}\} \\
Z &= \text{ENUM[no, neutral] + V} \\
S &= A \rightarrow N \rightarrow Z
\end{align*}
\]

\[
\begin{align*}
\text{VAR} \quad s : S := \{\ast\rightarrow\{\ast\rightarrow\text{neutral}\} \\
\quad \text{outcome} : A \rightarrow (V + \text{Null}) := \{\ast\rightarrow\text{nil}\} \\
\quad \text{allowed} : L \rightarrow \text{SET} V := \{\ast\rightarrow\{\}\}
\end{align*}
\]

\[
\begin{align*}
\text{% Agreement: } (\text{outcome.rng} – \{\text{nil}\}).\text{size} <= 1 \\
\text{% Validity: } (\text{outcome.rng} – \{\text{nil}\}) <= (\bigvee : \text{allowed.rng})
\end{align*}
\]

Paxos ensures that a round has a single value by having at most one leader process per round, and making the leader’s identity part of the round number. So \( N = \{i, l\} \), and leader \( l \) chooses \((i, l)\) for \( n \), where \( i \) is an \( i \) that has not used before, for instance, a local clock. The leader keeps \( v_n \) in a volatile variable; rather than resuming an old round after a crash, it just starts a new one.

To understand why the algorithm works, it’s useful to have the notion of a \text{stable} predicate on the state, a predicate that once true, remains true henceforth. Since the non-\( \text{nil} \) value of a round can’t change, \( v_n = v' \) is \text{stable}. Since a state value in an agent once set can’t change, \( s_n^a = v \) and are \text{stable}; hence “round \( n \) is dead” and “round \( n \) is successful” are stable as well. Note that \( s_n^a = \text{neutral} \) is not \text{stable}, since it can change to \( v \) or \( \text{no} \).

Here is a more complex \text{stable} predicate:

\[
\text{round } n \text{ is safe} = (\text{ALL } n' \mid n' <= n \Rightarrow n' \text{ is dead } \bigvee v_{n'} = v_n)
\]

In other words, if you look back at rounds before \( n \), skipping dead rounds, you see \( v_n \). If all preceding rounds are dead, any \( v_n \) makes \( n \) safe. For this to be well-defined, we need an ordering on \( v_n \’ s \), and for it to be useful we need a total ordering. We get this as the lexicographic ordering on the \( N = \{i, l\} \) pairs. This means that we must assume a total ordering on the leaders \( L \).

With these preliminaries, we can give the abstraction function from \text{Paxos} to \text{LateConsensus}. For allowed it is just the union of the leaders’ allowed sets. For outcome it is the value of a successful round.

\[
\text{ABSTRACTION FUNCTION}
\]

\[
\begin{align*}
\text{LateConsensus.allowed} &= \bigvee : \text{allowed.rng} \\
\text{LateConsensus.outcome} &= \{n \mid \text{Successful}(n) \mid \text{Value}(n).\text{choose}
\end{align*}
\]

\[
\begin{align*}
\text{FUNC} \quad \text{Successful}(n) \rightarrow \text{Bool} &= \text{RET} \{a \mid s_n^a \text{ IS } V, \text{majority} \\
\text{FUNC} \quad \text{Value}(n) \rightarrow (V + \text{Null}) &= \text{IF} \text{VAR } a, v \mid s_n^a = v \Rightarrow \text{RET } v [\ast] \text{ RET nil FI}
\end{align*}
\]

For this to be an abstraction function, we need an invariant:

(II) Every successful round has the same value.
It’s easy to see that this follows from a stronger invariant: If round \( n^* \) is successful, then any later round’s value is the same or nil.

\[(I2) \quad (\forall n', n \mid n' \leq n \land n' \text{ successful}) \implies v_n = \text{nil} \lor v_n = v_{n'}\]

This in turn follows easily from something with a weaker antecedent:

\[(I3) \quad (\forall n', n \mid n' \leq n \land n' \text{ is not dead}) \implies v_n = \text{nil} \lor (\forall n' \mid n' \leq n \implies n' \text{ is dead}) \lor v_n = v_{n'}\]

For validity, we also need to know that every round’s value is allowed:

\[(I4) \quad (\forall n \mid v_n = \text{nil} \lor v_n \text{ IN} (\land : \text{allowed}, \text{rng}))\]

Initially all the \( v_n \) are nil so that (I3) and (I4) hold trivially. The Paxos algorithm maintains (I3) by choosing the value of a round so that the round is safe. To accomplish this, the leader chooses a new \( n \) and queries all the agents to learn their state in all rounds with numbers less than \( n \).

Before an agent responds, it changes any neutral state for a round earlier than \( n \) to no.

Responses to this query from a majority of agents give the leader enough information to make round \( n \) safe, as follows:

It looks back from \( n \), skipping over rounds with no \( v \) state, since these must be dead (remember that the reported state is a \( v \) or no). When it comes to a round \( n^* \) with \( s_{n^*} = v \) for some agent \( a \), it chooses \( v \) as \( v_n \). Since \( n^* \) is safe by (I3), and all the rounds between \( n^* \) and \( n \) are dead, \( n \) is also safe.

If all previous rounds are dead, the leader chooses any allowed value for \( v_n \). In this case \( n \) is certainly safe.

Because ‘safe’ and ‘dead’ are stable properties, no state change can invalidate this choice.

Another way of looking at this is that because a successful round is not dead and can never become dead, it forms a barrier that the leader can’t get past in choosing a value that will make a later round safe. Thus the successful round forces the leader of the later round to choose the same value. A later round can choose differently from an earlier one only if the earlier one is dead.

An example may help to clarify the idea. Assume the allowed set is \( \{x, y, w\} \). Given the following two sets of states in rounds 1 through 3, with three agents \( a, b, \) and \( c \), the leader’s possible choices for round 4 are as indicated. In the middle column the leader can see a majority of agents in each round without seeing any \( v \)’s, because all the rounds are dead. In the right-hand column there is a majority for \( w \) in round 2.

<table>
<thead>
<tr>
<th>Leader</th>
<th>Message</th>
<th>Agent a</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Query</td>
<td>query(n) → for all n' &lt; n,</td>
</tr>
<tr>
<td></td>
<td>a</td>
<td>← report(a, s_{n'})</td>
</tr>
<tr>
<td>1</td>
<td>Command</td>
<td>command(n, v) ← if s_{n} = neutral then s_{n+1} := v</td>
</tr>
<tr>
<td></td>
<td>a</td>
<td>of agents to accept v</td>
</tr>
<tr>
<td>1</td>
<td>Agree</td>
<td>outcome(v) →</td>
</tr>
<tr>
<td></td>
<td>a</td>
<td>of a majority</td>
</tr>
</tbody>
</table>

When does Paxos terminate? If no leader starts another round until after an existing one is successful, then the algorithm definitely terminates as soon as the leader succeeds in both querying and commanding a majority. It doesn’t have to be the same majority for both, and the agents don’t all have to be up at the same time. Therefore we want a single leader, who runs one round at a time. If there are several leaders, the one running the biggest round will eventually
succeed, but if new leaders keep starting bigger rounds none may ever succeed. This is fortunate,
since we know from the Fischer-Lynch-Paterson result that there is no algorithm that is
guaranteed to terminate.

It’s easy to keep from having two leaders at once if there are no failures for a while, the
processes have clocks, and the maximum time to send, receive, and process a message is known:

Every potential leader that is up broadcasts its name.

You become the leader one round-trip time after doing a broadcast unless you have received
the broadcast of a bigger name.

The algorithm makes minimal demands on the properties of the network: lost, duplicated, or
reordered messages are OK. Because nodes can fail and recover, a better network doesn’t make
things much simpler. We model the network as a broadcast medium from leader to agents; in
practice this is usually coded by individual messages to each agent. We describe continuous
retransmission; in practice agents retransmit only in response to the leader’s retransmission.

A process acting as a leader uses messages to communicate with the same process acting as an
agent, so we describe the two roles of each process completely independently. In fact, the leader
need not be an agent at all.

The next section gives the algorithm in detail. You can skip this, but be sure to read the
following section on optimizations, which has important remarks about using Paxos in practice.

The Paxos algorithm: The details

We give a straightforward version of the algorithm in detail, and then describe encodings that
reduce the information stored and transmitted to small fixed-size amounts. The first set of types
and variables is copied from the earlier declarations.

```lisp
  MODULE Paxos[ % implements Consensus
    V, % data Value to agree on
    L WITH (*\<=>*: (L, L)\->Bool SUCHTHAT IsTotal), % Leader; <= a total order
    A WITH {majority: SET A->Bool} SUCHTHAT IsMaj ] % Agent

  TYPE I = Int
  N = [i, l] WITH {*\<<=*:LLeqN} % round Number; <= total
  Z = (ENUM[no, neutral] + V) % one agent’s state in one round
  A -> N -> Z % Agents’ states

  VAR outcome : A -> (V+Null) :=(*->nil) % the agents’ state, in
  s : S :=(*->neutral) % two parts.
  allowed : L -> SET V :=(*->\{\}) % The rest of the leaders’ state is the variables of LeaderActions(l).
  % All volatile except for n, which we make stable for simplicity.

  TYPE K = ENUM[query, command, outcome, report] % Kind of message
  M = [k, n, x: (Null + S + V) ] % Message; kind of message, % about round n

  Phase = ENUM[idle, querying, commanding] % defined for just one a % Actions for handling messages
  CONST n0 := N{i:=0, l:={l | true}.min} % smallest N

  % External actions. Can happen at any leader or agent.
  APROC Allow(v) = << VAR l | allowed(l) := allowed(l) \| {v} >>
  APROC Outcome() -> V = << VAR a | RET outcome(a) >>

  THREAD LeaderActions(l) =
    phase := idle; allowed := {}; reports := {}; v := nil
    phase = idle => VAR i | i > n.i => n.i := i; reports := {}; phase := querying
    phase = querying => Send(query, n, nil)
    phase = querying => VAR m := Receive(report) |
      s1 := m.x, a |
      reports := S{}; phase := commanding >>
    phase = commanding => Send(command, n, v)

  % New round. Hope n becomes largest N in use so this round succeeds.
  phase := idle \| i > n.i => n.i := i; reports := {}; phase := querying
  phase = querying => Send(query, n, nil)

  % Receive report: message for the current round.
  << phase = querying => VAR m := Receive(report) |
    m.n = n => reports := m; v := nil
    phase := idle; allowed := {}; reports := {}; v := nil

  % Start command. Note that Dead can’t be true unless reports has a majority.
  << phase = querying =>
    IF VAR nil |
      (ALL n' | n1 < n' \| n' < n => Dead(reports, n')) |
      m := m + | Val(reports, n1) | nil => v := Val(reports, n1)
    |
    phase := commanding >>

  % Send command message.
  phase = commanding => Send(command, n, v)

  % Receive report: message for the current round with non-neutral state.
  << phase = commanding => VAR m := Receive(report), sl := m.x, a |
    m.n = n \| sl1a \| sl2a \| neutral => reports := sl1a >>

  % Send outcome message if a majority has V state.
  Successful(reports, n) => Send(outcome, n, v)
```

% This round is dead if a majority has no state. Try another round.
Dead(reports, n) => phase := idle

---

FUNCTION SafeVals(n) -> SET V =

% The v’s that would make a safe round.

FUNCTION GoodVals(n) -> SET V =

% The good values for round n: if there’s no v yet, the ones that satisfy
% (13) and validity. If there is a v, then v.

FUNCTION ValAnywhere(n) -> SET V =

% Any trace of the value of round n, in messages, in s, or in a leader.

---

HANDOUT 18. Consensus
Reducing state and message sizes

It’s not necessary to store or transmit the complete agent state $s^a$. Instead, everything can be encoded in a small fixed number of $n$'s, $\lambda$'s, and $v$'s, as follows.

The relevant part of $s$ in an agent or in a report message consists of $v_{\text{last}}$ in some round last, plus no in all rounds strictly between last and some later round next, and neutral in any round after last and at or after next. Hence $s$ can be encoded simply as $(v_{\text{last}}, \text{last}, \text{next})$. The part of $s$ before last is just history and is not needed for the algorithm to run, because we know that last is safe. Making this precise, we have

$$\text{VAR } y: \{z, \text{last: } N, \text{next: } N\} := (\text{no, no, no})$$

and $s^a$ is just

$$\{ n \mid (n <= y.\text{last} \Rightarrow y.z [*] n < y.\text{next} \Rightarrow \text{no} [*] \text{neutral})\}$$

Note that this encoding loses the details of the state for rounds before last, but the algorithm doesn’t need this information. Here is a picture of this coding scheme.

<table>
<thead>
<tr>
<th>Round</th>
<th>$n_0$</th>
<th>. . .</th>
<th>last</th>
<th>. . .</th>
<th>next</th>
<th>. . .</th>
</tr>
</thead>
<tbody>
<tr>
<td>Agent state</td>
<td>some mixture of $v$'s and no</td>
<td>$v_{\text{last}}$</td>
<td>all no</td>
<td>neutral from here on</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

or just after accepting:

<table>
<thead>
<tr>
<th>last</th>
<th>. . .</th>
<th>next</th>
<th>. . .</th>
</tr>
</thead>
<tbody>
<tr>
<td>$v_{\text{last}}$</td>
<td>neutral from here on</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

In a leader, there are two cases for reports.

- If phase = querying, reports consists of the $s^a$'s, for rounds less than $n$, transmitted by a set of agents $a$. Hence it can be encoded as a set of “last state” tuples $(a, \text{last}_a, v)$. From this we care only about the number of $a$’s (assuming that $A$.majority just counts agents), the biggest last, and its corresponding $v$. So the whole thing can be coded as (count of $A$’s, $\text{last}_{\text{max}}$, $v$).

- If phase = commanding, reports consists of a set of $v_{\text{last}}$, or no in round $n$, so it can be encoded as the set of agents responding. We only care about a majority, so we only need to count the number of agents responding.

The leaders need not be the same processes as the agents. A leader doesn’t really need any stable state, though in the algorithm as given it has $n$. Instead, it can poll for the next’s from a majority after a failure and choose an $n$ with a bigger $n$. This will yield an $n$ that’s larger than any $n$ from this leader that has appeared in a command message so far (because $n$ can’t get into a command message without having once been the value of next in a majority of agents), and this is all we need to keep $s$ good. In fact, if a leader ever sees a report with a next bigger than its own $n$, it should either stop being a leader or immediately start another round with a new, larger $n$, because the current round is unlikely to succeed.

Combining rounds

In the most important applications of Paxos, we can combine the first round trip (query/report) with something else. For a commit algorithm, we can combine the first round-trip with the prepare message and its response; see handout 27 on distributed transactions.

The most important application is a state machine that needs to agree on a sequence of actions. We can number the actions $a_0, a_1, \ldots, a_k$, run a separate instance of the algorithm for each action, and combine the query/report messages for all the actions. Note that these action numbers, which we are calling $K$’s, are not the same as the Paxos round numbers, the $n$’s; each action has its own instance of Paxos and therefore its own set of round numbers. In this application, the leader is usually called the primary. The following trick allows us to do the first round trip only once after a new primary starts up: interpret the report messages for action $k$ as applying not just to consensus on $a_k$, but to consensus on all $\forall j > k$.

As long as the same process continues to be leader, it can keep track of the current $k$ in its private state. A new leader needs to learn the first unused $k$. If it tries to get consensus using a number that’s too small, it will discover that there’s already an outcome for that action. If it uses a number $k$ that’s too big, however, it can get consensus. This is tricky, since it leads to a gap in the action numbers. Hence you can’t apply the agreed action $k$, since you don’t know the state at $k$ because you don’t know all of the preceding actions that are applied to make that state. So a new leader must find the first unused $k$ for its first action $a$. A clumsy way to do this is to start at $k = 0$ and try to get consensus on successive $a_k$’s until you get consensus on $a_k$.

You might think that this is silly. Why not just poll the agents for the largest $k$ for which one of them knows the outcome? This is a good start, but it’s possible that consensus was reached on the last $a_k$ (that is, there’s a majority for it among the agents) but the outcome was not published before the leader crashed. Or it was not published to a majority of the agents, and all the ones that saw it have also failed. So after finding that $k$ is the largest $k$ with an outcome, the leader may still discover that the consensus on $a_k+1$ is not for its action $a$, but for some earlier action whose successful outcome was never broadcast. If leaders follow the rule of not starting consensus on $k+1$ until a majority knows the outcome for $k$, then this can happen at most once. It may be convenient for a new leader to start by getting consensus on a SKIP action in order to get this complication out of the way before trying to do real work.

Further optimizations are possible in distributing the actions already agreed, and handout 28 on primary copy replication describes some of them.

Note that since a state machine is completely general, one of the things it can do is to change the set of agents. So agents can be added or dropped by getting consensus among the existing agents. This means that no special algorithm is needed to change the set of agents. Of course this must be done with care, since once you have gotten consensus on a new set of agents you have to get a majority of them in order to make any further changes.
Leases

In a synchronous system, if you want to avoid running a full-blown consensus algorithm for every action that reads the state, you can instead run consensus to issue a lease on some component of the state. The lease guarantees that the state component won’t change (unless you have an exclusive lease and change it yourself) until some expiration time, a point in real time. Thus a lease is just like a lock, except that it times out. Provided you have a clock that has a known maximum difference from real time, you can be confident that the value of a leased state component hasn’t changed (that is, that you still hold the lock). To keep control of the state component (that is, to keep holding the lock), you can renew the lease before it expires. If you can’t talk to all the processes that have the lease, you have to wait for it to expire before changing the leased state component. So there is a tradeoff between the cost of renewing a lease and the time you have to wait for it to expire after a (possible) failure.

There are several variations:

- If you issue the lease to some known set of processes, you can revoke it provided they all acknowledge the revocation.
- If you know the maximum transmission time for a message, you can get by with clocks that have known differences in running rate rather than known differences in absolute time.
- Since a lease is a kind of lock, it can have different lock modes, for instance, ‘read’ and ‘write’.

The most common application is to give some set of processes the right to cache some part of the state, for instance the contents of a cache line or of a file, without having to worry about the possibility that it might change. If you don’t have a lease, you have to do an expensive state machine action to get a current result; otherwise you might get an old result from some replica that isn’t up to date. (Of course, you could settle for a result as of action \(k\), rather than a current one. Then you would need neither a lease nor a state machine action, but the client has to interpret the \(k\) that it gets back along with the result it wanted. Usually this is too complicated for the client.)

If the only reason for running consensus is to issue a lease, you don’t need stable state in the agents. If an agent fails, it can recover with empty state after waiting long enough that any previous lease has expired. This means that the consensus algorithm can’t reliably tell you the owner of such a lease, but you don’t care because it has expired anyway. Schemes for electing a leader usually depend on this observation to avoid disk writes.

You might think that an exclusive lease allows the process that owns it to change the leased state as well, as with ‘owner’ access to a cache line or ownership of a multi-ported disk. This is not true in general, however, because the state is replicated, and at least a majority of the replicas have to be changed. There’s no reliable way to do this without running consensus.

In spite of this, leases are not completely irrelevant to updates. With a lease you can use a simple read-write memory as an agent for consensus, rather than a fancier one that can do compare-and-swap, since the lease allows you do the necessary read-modify-write operation atomically under the lease’s mutual exclusion. For this to work, you have to be able to bound the time that the write takes, so you can be sure that it completes before the lease expires. Actually, the requirement is weaker: a read must see the atomic effect of any write that is started earlier than the read. This ensures that if the write is started before the lease expires, a reader that starts after the lease expires will see the write.

This observation is of great practical importance, since it lets you run a replicated state machine where the agents are ‘dual-ported’ disk drives that can be accessed from more than one machine. One of the machines becomes the master by taking out a lease, and it can then write state changes to the disks.

Compare-and-swap agents

An alternative to using simple memory and leases is to use memory that implements a compare-and-swap or conditional store operation. The spec for compare-and-swap is

\[
\text{APROC } \text{CAS}(a, \text{old}: V, \text{new}: V) \rightarrow V =
\begin{cases} 
\text{IF } \text{m}(a) = \text{old} \rightarrow \text{m}(a) := \text{new}; \text{RET old} [*] \\
\text{RET } \text{m}(a) 
\end{cases}
\]

Many machines, including the IBM 370 and the DEC Alpha, have such an operation (for the Alpha, you have to program it with Load Locked and Store Conditional, but this is easy and cheap).

To use a \text{CAS} memory as an agent, we have to code the state so that we can do the query and command actions as single \text{CAS} operations. Recall that the coded state of an agent is \(y = (v_{last}, \text{last}, n)\), representing the value \(v_{last}\) for round \(\text{last}\), \(\text{no}\) for all rounds strictly between \(\text{last}\) and \(n\), and \text{neutral} for all rounds starting at \(n\). A query for round \(n\) changes the state to \((v_{last}, \text{last}, n)\) provided \(\text{next} = n\). A command for round \(n\) changes the state to \((v_{n}, \text{last}, n)\) provided \(\text{next} = n\). So we need a representation that allows us to atomically test the current value of \(\text{next}\) and change the state in one of these ways. This is possible if an \(n\) fits in a single memory cell that \text{CAS} can read and update atomically. We can store the rest of the triple in a data structure on the side that is indexed by \(\text{next}\). If each possible leader has its own piece of this data structure, they won’t interfere with each other when they are updating it.

Since this is hard concurrency, the details of such a representation are tricky.

Complex updates

The actions of a state machine can be arbitrarily complex. For example, they can be complete transactions. In a replicated state machine the replicas must agree on the sequence of actions, and then each replica must do each action atomically. In handouts 19 and 20, on sequential and concurrent transactions, we will see how to make arbitrary actions atomic using redo recovery and locking. So in a general state machine each agent will write a redo log, in which each committed transaction corresponds to one action of the state machine. The agents must reach consensus on the complete sequence of actions that makes up the transaction. In practice, this means that each agent logs all the updates, and then they reach consensus on committing the transaction. When an agent recovers from a failure, it runs redo recovery in the usual way. Then it has to find out from other agents about any actions that they agreed on while it was down.
Of course, if several leaders are trying to run transactions at the same time, you have to make sure that the log entries don’t get mixed up. Usually this is done by funneling everything through a single master called the primary; this master also acts as the leader for consensus.

Another way of doing this is to use a single master with passive agents that just implement simple memory; usually these are disk drives that record redundant copies of the log. The previous section on leases explains how to run Paxos with such passive agents. When a master fails, the new one has to sort out the consensus on the most recent transaction as part of recovery.

**Batching**

Another way to avoid paying for consensus each time the state machine does an action is to batch several actions, possibly from different clients, into one super-action. They you get consensus on the super-action, and each replica can apply the individual actions of the super-action. You still have to pay to send the information that describes all the actions to each replica, but all the per-message overhead, plus the cost of the acknowledgements, is paid only once.