Optimistic Recovery in Multi-threaded Distributed Systems *

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Abstract

The problem of recovering distributed systems from crash failures has been widely studied in the context of traditional non-threaded processes. However, extending those solutions to the multi-threaded scenario presents new problems. We identify and address these problems for optimistic logging protocols.

There are two natural extension to optimistic logging protocols in the multi-threaded scenario. The first extension is process-centric, where the points of internal nondeterminism caused by threads are logged. The second extension is thread-centric, where each thread is treated as a separate process. The process-centric approach suffers from false causality while the thread-centric approach suffers from high causality tracking overhead. By observing that the granularity of failures can be different from the granularity of rollbacks, we design a new balanced approach which incurs low causality tracking overhead and also eliminates false causality.

1. Introduction

Multi-threading is becoming increasingly common in distributed systems owing to the need for light-weight concurrency. We address the problem of recovering multithreaded distributed systems from process crash failures. Although recovery has been a widely studied problem in traditional non-threaded systems [5], extending these solutions to the multi-threaded scenario gives rise to new problems. We address those problems for the optimistic logging protocols.

The traditional distributed recovery problem deals with recovering a distributed system from process crash failures. One approach to solving the recovery problem is log-based rollback recovery, which combines checkpointing and message logging. When a failure occurs, the distributed system can make use of the checkpoints and message logs to restore Vijay K. Garg Dept. of Electr. and Comp. Engg. Univ. of Texas at Austin garg@ece.utexas.edu

itself to a consistent global state. We focus on optimistic logging, an important class of log-based rollback recovery.

Optimistic logging protocols log messages to stable storage asynchronously, thus incurring low failure-free overhead. On a failure, some unlogged messages may be lost, resulting in the loss of some states of the failed process. Furthermore, this results in the rollback of states on other non-failed processes that causally depend on the lost states. In order to determine which states need to be rolled back, the causal dependencies between states needs to be tracked. This can be implemented by having all messages piggyback a dependency vector of size O(n), where n is the number of processes in the system [15].

While extending this solution to multi-threaded processes we have two natural choices: a *process-centric* approach and a *thread-centric* approach. In the processcentric approach, the internal non-deterministic events caused by threads are logged [7, 14]. With this provision, other researchers have used traditional optimistic protocols. This, however, gives rise to the problem of false causality between threads of a process. This problem has two serious repercussions. First, during failure-free mode, it causes the unnecessary blocking of outputs to the environment. Second, during recovery from a failure, it causes unnecessary rollbacks.

Attempting to eliminate false causality leads to the thread-centric approach. Here, each individual thread is treated as a process and a process crash is treated as multiple concurrent thread crashes. In this approach, during failure-free operation, causality is tracked at the level of threads. This makes causality tracking an expensive operation requiring a dependency vector of size O(nm), where *n* is the number of processes and *m* is a bound on the number of threads per process. This increases the message size overhead, as well as space and time overhead.

Thus the process-centric and the thread-centric approaches present a trade-off between false causality and tracking overhead. We make the observation that processes fail independently and are thus *failure units* and that threads may be rolled back independently and are thus *rollback*

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units. Therefore, it is sufficient to track the dependency of threads on processes. This *balanced* approach tracks causal dependencies using a dependency vector of size O(n). At the same time it eliminates false causality, since threads are rolled back independently.

In Section 2, we present some background on optimistic recovery in traditional non-threaded environments. In Section 3, we describe the two natural extensions – process-centric logging and thread-centric logging – and the associated false causality versus tracking overhead trade-off. In Section 4, we describe our new *balanced protocol* for optimistic recovery in multi-threaded distributed systems. Section 5 is a note on generalizing the ideas of the paper.

2. Background: Optimistic Recovery

2.1. System Model and the Recovery Problem

We consider an application system consisting of n processes communicating only through messages. The communication system used is unreliable, in that it can lose, delay, or duplicate a message. The environment also uses messages to provide inputs to and receive outputs from the application system. Each process has its own volatile storage and also has access to stable storage. The data saved on volatile storage is lost in a process crash, while the data saved on stable storage remains unaffected by a process crash.

A process execution is a sequence of states. The states may be divided into *state intervals* consisting of the states between two consecutive message receipts by the application process. In single threaded systems, the execution within each interval is assumed to be completely deterministic, i.e., actions performed between two message receives are completely determined by the content of the first message received and the state of the process at the time of the first receive. As we will see, in multi-threaded systems, non-deterministic thread scheduling affects the state of a process. It is sufficient for our purposes to view process executions at the granularity of state intervals and not states. Therefore, for simplicity, we will sometimes use *states* to mean *state intervals*.

All n process executions together constitute a system execution. Two physical system executions are considered equivalent if their interaction with the environment is the same.

A process fails by simply crashing. In a crash failure, a process stops executing and loses the data in its volatile storage. The process does no other harm, such as sending incorrect messages. Pre-failure states of a process that cannot be recreated after a failure are called lost states.

The application system is controlled by an underlying recovery system. The type of control may be of various forms, such as saving a checkpoint of the application process, stopping an application process, adding control information to the state of an application process, adding control information to a message, rolling back the application to an earlier state, etc.

The *recovery problem* is to specify the behavior of a recovery system that controls the application system to ensure that despite crash failures, the system execution remains equivalent to a possible crash-free execution of the standalone application system.

2.2. Optimistic Logging

Log-based rollback recovery protocols [5] rely on checkpoints and message logs, using them during recovery to restore the whole system to a consistent global state (one in which every received message was sent). It is guaranteed that this restored state is one which could possibly have happened in a failure-free system execution and, therefore, this approach solves the recovery problem. Depending on when and where the received messages are logged, the log-based rollback recovery schemes can be divided into three categories: pessimistic, optimistic, and causal [5]. In this paper, our focus is on optimistic logging protocols.

We first present an example and then use it to specify the details of how a traditional optimistic logging protocol operates. The protocol we present is similar in spirit to the ones presented in [3, 15].

Example

An example of an optimistic recovery system is shown in Figure 1. Solid horizontal lines show the useful computation, and dashed horizontal lines show the computation that is either lost in a failure or rolled back by the recovery protocol. In the figure, c1 and c2, shown by squares, are checkpoints of processes P1 and P2 respectively. State intervals are numbered from s0 to s7 and they extend from one message receive to the next. The numbers shown in rectangular boxes will be explained later in this chapter.

In Figure 1(a), process P1 takes a checkpoint c1, acts on some messages (not shown in the figure) and starts the interval s0. P1 logs to stable storage all messages that have been received so far. It starts interval s2 by processinging the message m0. In interval s2, message m2 is sent to P2. P1then fails without logging the message m0 to stable storage or receiving the message m1. It loses its volatile memory, which includes the knowledge about processing the message m0. During this time, P2 acts on the message m2.

Figure 1(b) shows the post-failure computation. On restarting after the failure, P1 restores its last checkpoint c1, replays all the logged messages and restores the interval s1. It then broadcasts a failure announcement (not shown in Figure 1). It continues its execution and starts interval s6 by processing m1. P2 receives the failure announcement in interval s5 and realizes that it is dependent on a lost state. It rolls back, restores its last checkpoint c2, and replays the

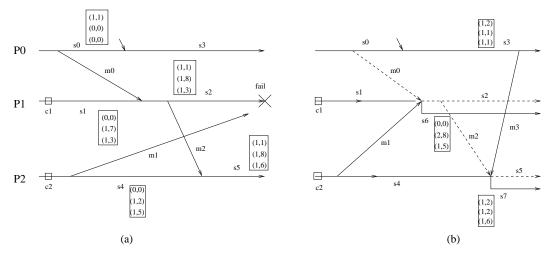


Figure 1. Example: Optimistic Recovery in Action

logged messages until it is about to process m2, the message that made it dependent on a lost state. It discards m2and continues its execution by processing m3. The message m2 is not regenerated in post-failure computation. P0 remains unaffected by the failure of P1.

Detecting Orphans: Causally Precedes

As just seen, in optimistic logging, some messages may be lost in a failure. This may result in some lost states on the failed process. All states that "causally depend" on such lost states must also be detected and rolled back. Such states are known as *orphans*.

The intuitive notion of "causally depends" that we have used is formalized by the following relation. Let *causally precedes* (denoted by \rightarrow) be the smallest transitive binary relation on state intervals satisfying the following two conditions:

- u → v if the processing of an application message in state u results in state v, (for example, s1 → s6 in Figure 1(b)),
- u → v if the processing of an application message sent from u starts v (for example, s2 → s5 in Figure 1(a)).

By $s \rightarrow u$, we mean $s \rightarrow u$ or s = u. Note that a failure or a rollback does not start a new interval. It simply restores an old interval. Now we can define an orphan state as:

$$orphan(s) \equiv \exists u : lost(u) \land u \rightarrow s$$

Tracking Causal Dependencies: Dependency Vectors

The causally precedes relation needs to be tracked for orphan detection. For this purpose, a *dependency vector* (dv) is used. The dv of a process P_i has *n* state interval indices, where *n* is the number of processes in the system and a state interval index is a tuple containing an incarnation number and a sequence number.

Whenever a process fails and is restarted, it is said to be in a new *incarnation*. The incarnation number in the *i*'th entry of its dv is its own incarnation number. The incarnation number in the *j*'th entry is equal to the highest incarnation number of P_j which causally precedes P_i . Let state interval index e be (t, i). Then, we define a total ordering, <, on state interval indices as $e_1 < e_2 \equiv (t_1 < t_2) \lor [(t_1 = t_2) \land (i_1 < i_2)].$

Each process piggybacks its dv on every outgoing message. Before processing a message, a process updates its dv by taking a componentwise maximum of its dv with the dv of the incoming message and incrementing its own sequence number.

An example of dv is shown in Figure 1. The dv of each state is shown in a rectangular box near it. The k'th row of the rectangular box corresponds to dv[k].

It has been previously demonstrated that dependency vectors track the causally precedes relation, and therefore, can be used to detect orphans [3, 15].

Recovering from a Crash

When a process P_j fails, it restores its most recent checkpoint and replays the logged messages that were processed after that checkpoint. Next, P_j broadcasts a failure announcement containing its state index, which is the ending index number of the failed incarnation. In Figure 1, failure announcement of P_2 contains (1,7). It then waits for an acknowledgment from all processes.

Upon receiving a failure announcement, a process P_i compares its dv with that index. If the dv shows that P_i 's

state depends on a higher-index interval of the failed incarnation of P_j , P_i rolls back to undo the orphan states. Similarly, it discards the orphan messages from its log. It also saves the received failure announcement in an *incarnation end table* to discard any orphan message that may arrive in future. It then sends an acknowledgment to the sender of the failure announcement.

Handling Output Commits

Distributed applications often need to interact with the outside world. Examples include setting hardware switches, performing database updates, printing computation results, displaying execution progress, etc. Since the outside world in general does not have the capability of rolling back its state, the applications must guarantee that any output sent to the outside world will never need to be revoked. This is called the *output commit problem*.

In optimistic recovery, an output can be committed when the state intervals that the output depends on have all become *stable* [15]. (An interval is said to be *stable* if it can be recreated from the information saved on stable storage). To determine when an output can be committed, each process periodically broadcasts a logging progress notification to let other processes know which of its state intervals have become stable.

3. Optimistic Recovery with Multi-threaded Processes

3.1. System Model

Each process contains a set of threads and a set of shared objects. Threads of different processes communicate only through messages. Threads of the same process communicate through shared objects and messages. Any other form of communication is allowed between threads, as long as it can be modeled using shared objects or messages. For example, wait-notify synchronizations can be modeled using messages.

Threads of a process crash together. This happens not only in hardware crashes but also in most software crashes because threads share the same address space and are not protected from each other by the operating system.

The recovery system can restore the state of an individual thread or shared object to an old state without affecting the other threads or shared objects. This assumption will be discussed in Section 4.4.

3.2. Extending Optimistic Recovery

We now investigate how to extend the optimistic recovery protocol given in Section 2 to multi-threaded environments.

Strom and Yemini [15] presented the original optimistic protocol not in terms of processes, but in terms of *recovery*

units. A recovery unit is a single unit of execution in optimistic recovery systems. Recovery units fail as a unit and roll back in response to another unit's failure.

In previous sections, we chose individual processes as recovery units. In a multi-threaded environment, there are two natural candidates for the recovery unit: a process or a thread.

3.2.1 Process-centric Logging

In treating a process as a recovery unit in a multi-threaded system, there is another source of non-determinism apart from the order of message receives. Depending on the scheduling, the threads may access shared objects in a different order. Therefore, after a failure, replaying the message log to a process is not sufficient to recreate the desired states.

To solve this problem, Goldberg et. al. [7] require that shared objects be accessed only in locked regions. The order in which threads acquire locks is logged. During a replay, the same locking order is enforced. This trace-andreplay technique has also been used in concurrent debuggers [10, 16].

Another approach has been used by Elnozahy and Slye [14]. They focus on uniprocessor multi-threaded environments in which the points of non-determinism can be reduced to the thread switches. Therefore, they log the order of thread switches and ensure that thread switches occur in the same order during replay. Again, this approach has been used in concurrent debuggers [11, 13].

Given that the non-determinism due to thread scheduling can be tracked and replayed, the general optimistic recovery approach described before can be used with a process as a recovery unit.

The False Causality Problem

An example of how the process-centric approach operates is shown in Figure 2(a). Receive of message m1, m2 and m3 starts the intervals s1, s2 and s3 respectively. When P0fails and loses state interval s1, P1 has to roll back state interval s3. In the figure, the threads in each process are not shown, since processes are the recovery units.

Let us now take another look at this scenario at the level of threads instead of processes. Figure 2(b) shows the same scenario at the level of threads. The figure shows that process P1 consists of two threads T2 and T3. Therefore, s3is an interleaving of the states of T2 and T3. Suppose that, after m3 was received, there was no shared object interactions between T2 and T3. So, only the states on T3 were really caused by m3 and needed to roll back. The states in s3 belonging to T2 were rolled back unnecessarily. This is

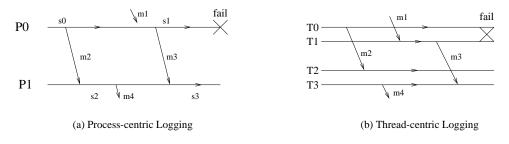


Figure 2. Extending Optimistic Logging

said to be due to *false causality* induced between the states in the two threads.

Besides causing unnecessary rollbacks during recovery, false causality has an unwanted effect in failure-free mode as well. In Figure 2(a), the output message m4 from s2 cannot be committed until s0 has become stable. However, the thread view in Figure 2(b) shows us that this was, in fact, unnecessary. The waiting was a result of false causality induced between m2 and m4. Thus, false causality also increases the latency of output commits. As seen in the previous section, the latency of output commits is an important factor for message logging protocols.

How often does false causality arise? Lewis and Berg [9] have divided multi-threaded programs into two main categories: inherently multi-threaded programs and not obviously multi-threaded programs. Inherently multi-threaded programs require multi-threading for ease of programming and not for speedup. These programs have highly independent tasks that are naturally expressed as threads. Some examples of such programs are: servers which handle multiple requests simultaneously, debuggers which monitor a program while keeping multiple displays active at the same time, and simulators which simulate different entities that operate simultaneously. The other class of not obviously multi-threaded programs are those that require multi-threading for speedup on a multi-processor machine. Such programs have tightly-coupled threads that interact frequently through shared memory. Some examples are numerical programs and fine-tuning bottlenecks in existing code. Of these two categories, the inherently multi-threaded programs have highly independent threads which do not interact frequently, and therefore, would display false causality more often. Our focus will be on this important category of multi-threaded applications.

Given that the false causality problem is an important concern, how can it be addressed? The problem arises because threads, which are independent units, are grouped together as a single unit. To solve this problem, we now study an approach that models each thread as a recovery unit.

3.2.2 Thread-centric Logging

Threads can be modeled as recovery units since they can be rolled back independently. Failure of a process is modeled as concurrent multiple failure of all its threads.

There are a number of important issues that arise when treating threads as recovery units in the optimistic logging scheme. First, in addition to dependencies between threads due to messages, there are also dependencies caused by shared objects. These new dependencies must be tracked. Second, on a failure, just as a thread may have to roll back, a shared object may also have to roll back. Thus, orphan detection must be carried out for threads, as well as for shared objects. Third, both threads and shared objects must be restored to a checkpoint and replayed.

In order to address these issues, we now describe a way to model shared objects using messages and threads. A fictitious thread is associated with each shared object whose state is the same as that of the corresponding object. We model each method invocation on a shared object as a pair of messages between the invoking thread and the thread associated with the object. The first message is sent by the invoking thread and contains the method identifier and the method parameters. The second message is sent by the thread associated with the object and contains the return value of the method. This greatly simplifies presentation of the protocols because now messages are the only form of communication in the whole distributed system.

Further, the simplified model captures a way to deal with each of the three issues mentioned above. The new dependencies are tracked by treating the shared object accesses as messages and associating a vector with each shared object. Similarly, orphan detection and replay of shared objects is done just as in threads.

Since we have dealt with the new issues, all that remains is to apply the general optimistic logging scheme. So the computation in Figure 2(a) appears as that in Figure 2(b). When T0 and T1 fail, the thread-interval sending m2 is not lost in the failure. So only T3 and not T2 is rolled back. Also, message m4 is committed without waiting for the interval sending m2 to become stable.

Therefore, the thread-centric approach clearly reduces

<pre>// and sequence numbers</pre>				
Thread T _i	Process P _i			
dv array[n] of sii_type; // dependency vector IET array[n] of set of sii_type; // incarnation end table	sii_type; // state interval index log list of untyped objects; // log for messages and sii LOG list of untyped objects; // stable log for messages, // sii values, and checkpoints			

Figure 3. Variables Used in the Protocol

false causality. The dependency tracking overhead, however, is greatly increased. A main factor in this overhead is that, instead of O(n) entries, each dependency vector now has O(mn) entries (where *m* is the maximum number of threads per process). A more detailed discussion on this overhead will be presented in Section 4.3.

(*inc*: int, *seq*: int);

An Inherent Trade-off?

type

sii_type:

The process-centric and thread-centric approaches offer a trade-off between dependency tracking overhead and extent of false causality. This trade-off seems to be an inherent one as it arises from the choice of granularity of the recovery unit. A larger recovery unit introduces more false causality and has lower tracking overhead than a smaller one. In database systems, an analogous trade-off exists between lock maintenance overhead and extent of false causality while choosing the lock granularity. Surprisingly, in multi-threaded recovery, this trade-off can be avoided by a scheme that we now present.

4. The Balanced Protocol

We observe that a recovery unit plays two distinct roles in optimistic recovery. The first role is that of a *failure unit*. The defining characteristic of a failure unit is that it fails as a unit. The second role is that of a *rollback unit*. A rollback unit can be rolled back and restored to a previous state independently. For example, in the process-centric protocol, the process was both the failure unit and the rollback unit, whereas in the thread-centric protocol, the thread was the failure unit and rollback unit.

A general observation we can make about optimistic recovery is that: *to detect orphans, it is sufficient for a rollback unit to track its transitive dependency on a failure unit.* Then, the failure of a failure unit causes all orphaned rollback units to rollback, bringing the system back to a consistent state.

Thus, choosing a larger granularity failure unit reduces the dependency tracking overhead since there are fewer entities to track. Also, choosing a larger granularity rollback unit increases the extent of false causality since multiple entities are forced to roll back together. In the previous section, we saw that the trade-off between dependency tracking overhead and false causality depended on the granularity of the recovery unit. The separation of roles into failure units and rollback units allows the trade-off to be avoided, by choosing a larger granularity failure unit (process) and a smaller granularity rollback unit (thread). This is the central idea for our *balanced protocol*.

// type representing state interval index with incarnation

However, to achieve the separation of roles, we must simultaneously deal with both thread state intervals and process state intervals. This requires a redefinition of a process state interval as a "consistent" set of thread state intervals. We define this notion of consistency formally in [4].

4.1. The Balanced Protocol: Details

The protocol specifies the actions to be taken by the recovery system. The actions are divided into two categories, those for a process (failure unit) and those for a thread (rollback unit).

Figure 3 shows the variables used in the protocol. We use capital letters to indicate variables on stable storage (e.g. IET) and small letters to indicate variables on volatile storage (e.g. dv). Global variables common to all threads (process variables) are underlined (e.g. <u>sii</u>). Calls to the run-time environment start with an underscore (e.g. $_Send(data, dv)$ calls the environment, while Send(data) calls the program's function).

Each thread maintains a dependency vector dv and an incarnation end table IET in order to detect orphans. The <u>sii</u> indicates the current process-wide state interval index. The <u>log</u> and <u>LOG</u> are the volatile and stable logs respectively. The volatile log is used as a buffer for the log before it is made stable. The stable log is used to recover from a process failure. The stable and volatile logs are also used to independently restore a thread to a previous state.

The actions of the protocol may be divided into two types: normal mode and recovery mode. Figure 4 lists the

Thread T_j	Process P _i	
Initialize: $\forall k \neq j : dv[k] := (0, 0)$ dv[j] := (1, 1) $\forall k : IET[k] := \{\}$	Initialize: $\frac{sii}{log} := (1,1)$ $\frac{log}{LOG} := null$	
Send($data$): _Send($data, dv$);	Take Checkpoint: $\frac{log := log + sii}{LOG := LOG} + log;$	
Receive: repeat $m := _Receive();$ until (\neg orphan($m.dv, IET$));	$\frac{log := null;}{\underline{LOG} := \underline{LOG} + _Checkpoint();}$ $\underline{sii}.seq := \underline{sii}.seq + 1;$	
	Make Message Log: $\underline{LOG} := \underline{LOG} + \underline{log};$ $\underline{log} := \text{null};$	
	Start State Interval: $\frac{log := log + sii}{\underline{sii}.seq := \underline{sii}.seq + 1};$	



normal mode protocol. For simplicity, all actions are assumed to be atomic.

First, in order to ensure that process state intervals are consistent, the recovery system periodically starts a new state interval (**Start State Interval**) by incrementing the global <u>sii</u> value. The old <u>sii</u> is queued in the log marking the end of the previous receive set in the log. On the next receive event, each thread T_j assigns its local dv[j] entry the value of this global <u>sii</u> (**Receive**), thus keeping track of the process state interval it belongs to.

Each thread keeps track of the highest process state interval that it is aware of using its local dependency vector dv. The dependency vector mechanism is the same as before. (Send, Receive).

On receiving a message, a thread must discard it, if it is an orphan message (**Receive**). It can detect this by looking at its incarnation end table IET which is updated appropriately in the recovery mode of the protocol. More precisely, we specify the predicate

$$orphan(dv, IET) = \exists j : \exists (t, x) \in IET[j] : \\ (t = dv[j].inc) \land (x < dv[j].seq)$$

A received message is logged in the global volatile log *log* (**Receive**). This log totally orders all receives of all threads in their real time order, marking the end of receive sets by storing their <u>sii</u> values (**Start State Interval**). Periodically, this volatile log is flushed to stable storage (**Make Message Log**).

Periodically, checkpoints are also taken. The volatile log is flushed to the stable log and a checkpoint is appended (**Take Checkpoint**). The old <u>sii</u> value is logged before the checkpoint and incremented after the checkpoint. This ensures that every checkpoint is exactly between two state intervals. We assume that in addition to the application system state, the checkpoint includes the dependency vector dvfor each thread.

The recovery mode protocol is listed in Figure 5. On a crash failure, the crashed process restores its last checkpoint in the stable log, <u>LOG</u> (**Restart After Crash**). All threads are replayed from this point using messages and <u>sii</u> values from <u>LOG</u> upto the last complete process state interval.

Next, the crashed process broadcasts an announcement of its failure to all threads of all other processes. This announcement includes its recovered state interval index indicating the end of that incarnation. This broadcast must be reliable in order to ensure that the system returns to a consistent state. Reliability may be ensured by repeating the broadcast periodically. The process then blocks, waiting for an acknowledgement from all threads. Once all of these are received, it starts a new incarnation by appropriately updating its state interval index.

When a thread receives a failure announcement, it first records the announcement in its incarnation end table *IET* (Handle Fail_Announce). This will be later used in normal mode to discard orphan messages (**Receive**). It then decides if it is an orphan based on its dependency vector

Thread T_j	Process P _i		
$\begin{array}{l} \textbf{Handle Fail_Announce}(sender_id, sender_sii):\\ IET[sender_id] \coloneqq IET[sender_id] \cup \{sender_sii\};\\ \text{if (orphan(}dv, IET))\\ \text{delete_orphans}(log, \underline{LOG}, IET)\\ \text{restore last checkpoint of } T_j \text{ in } \underline{LOG};\\ \text{locally replay thread } T_j \text{ using messages and } \underline{sii}\\ \text{values after last checkpoint of } T_j \text{ in } \underline{LOG} \text{ and } \underline{log};\\ \text{send } Ack_Announce \text{ to } P_{sender_id} \end{array}$	Restart After Crash: restore last checkpoint in <u>LOG</u> ; globally replay all threads in P_i using messages and <u>sii</u> values after last checkpoint in <u>LOG</u> ; reliably broadcast Fail_Announce(i, <u>sii</u>) to all threads in all other processes; wait for Ack_Announce from all; <u>sii</u> .inc := <u>sii</u> .inc + 1; <u>sii</u> .seq := 1;		

Figure 5. Protocol for Recovery-mode Operation

and its newly updated IET. If it is, it must delete all orphan entries from its stable and volatile logs. Next, it must restore the last thread checkpoint from the stable log. It then replays itself to its latest state using the messages and <u>sii</u> values in the stable and volatile logs. This will bring it to the latest state that is not an orphan with respect to the received failure announcement. Note that the other threads and, in particular, the global <u>sii</u> remain unaffected by this action. Finally, it sends an acknowledgement to the sender of the failure announcement.

To complete the protocol, we must add logging progress notification to accomodate output commit. However, we omit these details because they are identical to those in traditional optimistic protocols.

4.2. An Example

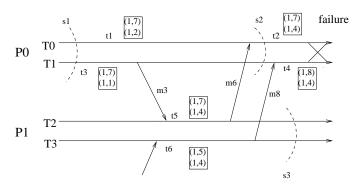


Figure 6. Example: Balanced Protocol

An example of our protocol in action is shown in Figure 6. Threads T0, T1 belong to process P0 and T2, T3 belong to P1. The dashed arcs show the ends of the process state intervals s1, s2 and s3. Thread state intervals are t1 to t6. State interval indices of s2 and s3 are (1,7) and (1,4) respectively. When P0 fails, it loses the state interval s2. It broadcasts a failure announcement containing the index (1,6), corresponding to the state interval index of s1. On receiving this announcement, thread T2 rolls back the

orphan state interval t5. Thread T3 remains unaffected by this failure. Note that, if P1 were to fail instead of P0, and lose the state interval s3, then both T0 and T1 will detect that they are orphans due to the entry (1,4) in their dv's. This illustrates an important point of our protocol: in spite of belonging to the same process state interval and sharing a common index, thread intervals t5 and t6 act as independent rollback units and a single failure unit at the same time.

4.3. Comparative Evaluation

There are two factors of interest while comparing various protocols: false causality and dependency tracking overhead.

We have already discussed the false causality problem in Section 3.2.1. To summarize: the false causality problem arises in the process-centric approach because threads are forced to roll back together even when they have low interactions between them. False causality is particularly a problem for a large class of applications that have low interactions between threads. The observable effects of false causality are: (1) delayed output commits, and (2) unnecessary rollbacks after a failure. Both the thread-centric and balanced approaches avoid false causality by allowing threads to roll back independently.

The price paid for avoiding false causality is the higher dependency tracking overhead. This overhead is in three forms: space overhead, time overhead, and message size overhead. Table 1 summarizes the relative overheads of the various protocols. The overhead of checkpointing is common to all protocols and hence it is not shown in the table.

As discussed before, there have been two implementations of the process-centric approach: Slye & Elnozahy [14], and Goldberg et al. [7]. Slye & Elnozahy use a software counter to track the thread switches. Therefore, the space overhead consists of O(s) space to log all thread switching information and O(en) space to store dependency vectors for each receive event. The time overhead consists of the total extra time the recovery protocol requires to execute. This involves the time to save check-

	Space Overhead	Time Overhead	Message Size Overhead	False Causality
Process-centric(I)	O(s + en)	O(c + en)	O(n)	yes
Process-centric(II)	O(o + en)	O(o + en)	O(n)	yes
Thread-centric	O(mn(o+e))	O(mn(o+e))	O(mn)	no
Balanced	O(n(o+e))	O(n(o+e))	O(n)	no

Process-centric(I) is the process-centric protocol using Slye & Elnozahy [14] Process-centric(II) is the process-centric protocol using Goldberg et al. [7] n is the number of processes

m is the maximum number of threads and shared objects per process

e is the maximum number of message receive events per process execution

o is the maximum number of shared object accesses per process execution

s is the maximum number of thread switches per process execution

c is the time overhead for maintaining a software counter

Table 1. Comparative Evaluation of Overheads

points, log thread switching information, log dependency vectors. Therefore, the time overhead is proportional to the space overhead. The message size overhead is O(n) since the dependency vector has n entries, one per process.

Goldberg et al. log the order of shared memory accesses so that they can be deterministically replayed. Therefore, the space overhead is O(o + en) with the O(o) component accounting for the log made on each shared memory access. The time overhead is proportional to space overhead. The message overhead remains O(n) as before.

For the thread-centric approach, we assume that the overheads of checkpointing are similar for thread and process checkpoints. In practice, thread checkpoints may take additional time overhead to separate the thread local state from the process address space. Another method would be to simply take process checkpoints and extract the thread checkpoints when required. Since shared object accesses are treated as message receives, the space overhead to log the dependency vectors is O(mn(o + e)) since each vector has O(mn) entries. The time overhead is similar. The message size overhead is now O(mn).

The balanced approach reduces the vector size from O(mn) in the thread-centric approach to O(n). All overheads are similar to the thread-centric case replacing mn by n.

The saving in space, time and message size overhead of the balanced protocol with respect to the thread-centric protocol is substantial because mn is potentially a very large quantity compared to n. Each individual thread and shared object in the system is accounted for in mn. Since both protocols achieve the same elimination of false causality, the balanced protocol should always be preferred to the threadcentric protocol.

Compared to the process-centric protocol of Goldberg et al., the balanced protocol has the same message size overhead, but higher space and time overhead. This is because each shared object access in the balanced approach logs a O(n) vector instead of constant information. With respect to the process-centric protocol using Slye & Elnozahy's technique, the space and time overhead is also expected to be higher since there are usually much fewer thread switches than shared memory accesses. However, as in Section 3.2.1, only the applications that have low thread interaction suffer greatly from false causality. For these applications, the increase in time and space overheads of balance protocol is low because the number of shared object accesses is low. Thus, the process-centric protocol should be used for applications with high thread interaction, and, therefore, low false causality effects. The balanced protocol should be used for the class of applications that have low thread interaction, where the extra space and time overhead is outweighed by the saving in false causality effects.

4.4. Implementation Issues

There are two new issues that arise when implementing optimistic recovery in multi-threaded systems as opposed to traditional systems: checkpointing threads, tracking causality through shared object accesses, and replaying threads independently.

Checkpointing threads is handled differently in different multi-threaded systems. In POSIX Threads, the thread state consists of the stack, register context (including program counter), thread local storage, and objects that are created on the heap by a thread. The global data is shared between all threads. This division may be simplistic, and it is possible for shared objects to exist in the stack of one process or on the heap. In such cases, a simplistic division would work correctly but would induce greater false causality. It is possible to track ownership of objects on the heap at the execution point when the object was created (e.g. by instrumenting at the *malloc* calls). It is important to note that the checkpoint can be made for the whole process as in existing schemes for multi-threaded systems [7]. This logically corresponds to all threads checkpointing at the same time. However, when required, the thread state must be extracted from the process checkpoint independently. In Java, object serialization allows the objects to be saved independently. The thread stacks are not exposed by the Java VM. In order to save thread stacks, existing schemes either modify the Java VM [12], or instrument the Java code [6].

An important issue in implementing shared objects is defining what exactly a shared object access is. This has impact on the number of accesses, *o*. As we have just seen, the smaller the value of *o*, the lower would be the overhead in the balanced protocol. Frequently, multi-threaded programs are written using a discipline that ensures that each shared object is accessed only through methods and all the methods are made mutually exclusive using an object lock. This is the model used by Java threads [8]. These mutually exclusive methods increase the granularity of a shared object access and provide a clear interface for tracking. Thus, instead of tracking every read and write to shared objects, we can now track each method invocation on a shared object.

Each method invocation on a shared object is treated as a message from the thread to the shared object and the return from the shared object is treated as a message from the shared object to the thread. Thus, the recovery manager must intervene at both these points. The log consists of the method and parameter values in case of the invocation and the returned values in the case of the return. While replaying a thread or a shared object, the logged values are replayed so that the thread or shared object recovers its state.

5. Final Note: Generalization

So far, we have considered the failure and rollback units in the context of processes and threads. We can select these units at even coarser granularities. For simplicity, let us consider a system consisting of non-threaded processes. Now, a process can be a rollback unit while a processor can be a failure unit. Compared to traditional optimistic protocols, this approach reduces the dependency tracking overhead while injecting false causality between processes. In case of a hardware crash, entire processor indeed crashes as a unit. In case of a software crash of a single process, all processes on the corresponding processor have to simulate a crash, and hence the false causality between them. Further generalizing, a local area network can act as a failure unit in a wide area environment. A process still acts as a rollback unit.

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