Execution replay and debugging*

Michiel Ronsse¹, Koen De Bosschere¹, and Jacques Chassin de Kergommeaux²

¹ELIS, Ghent University, St.-Pietersnieuwstraat 41, B-9000 Ghent, Belgium
²ID-IMAG, B.P. 53, F-38041 Grenoble Cedex 9, France

Abstract

As most parallel and distributed programs are internally non-deterministic – consecutive runs with the same input might result in a different program flow – vanilla cyclic debugging techniques as such are useless. In order to use cyclic debugging tools, we need a tool that records information about an execution so that it can be replayed for debugging. Because recording information interferes with the execution, we must limit the amount of information and keep the processing of the information fast. This paper contains a survey of existing execution replay techniques and tools.

1 Introduction

Conceptually, a parallel or distributed program¹ consists of a set of cooperating processes that are executed in parallel by different processors. Writing parallel programs is generally considered more difficult and more error-prone than writing sequential programs as one has not only to concentrate on the

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¹For the remainder of this paper, we will use the term ‘parallel program’ for a program consisting of a number of processes running on a multiprocessor with shared memory and the term ‘distributed program’ for a program running on a number of computers without shared memory.

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implementation of a particular algorithm, but also on the communication and synchronisation between the processes. Most contemporary parallel programming tools are focused on automatic parallelisation of sequential programs, or the analysis and visualisation of parallel programs, not on the development of hand-written parallel or distributed programs. Consequently, there is a clear lack of development tools for parallel and distributed programs.

A standard debugger that is used for ‘cyclic debugging’, i.e. re-executing the program over and over again with the same input, and zooming in on the program execution until the bug is found (using break points, watch points, etc.) assumes that a program can be re-executed deterministically. This is clearly not the case for many parallel and distributed programs. Indeed, a standard source level debugger will change the timing of the processes, and hence maybe alter the program flow. As a result, it can cause the symptoms of a particular bug to suddenly disappear, or to be replaced by other symptoms (so-called Heisenbugs [21]).

In order to be able to use the wealth of debugging tools that have been developed for sequential programs for the debugging of parallel programs, we need a way to deterministically re-execute a parallel program. The main problem is that parallel programs are non-deterministic (especially the faulty ones): each program run (even with the same input) might result in a different program execution. Although non-determinism is also present in sequential programs (caused by interrupts, etc.) its presence is far more visible in parallel programs because a lot of non-deterministic constructs are used on purpose: e.g., the order in which processors use a semaphore is not planned by the program code but is determined by the competition between the different processes.

One way to enable cyclic debugging techniques for parallel or distributed programs is by the usage of the so-called execution replay technique. This technique consists of two phases: first a trace of a parallel execution is made (record phase), and afterwards the trace is used to control the re-execution of the program (replay phase)\(^2\), provided one can supply the same input to the program as the one supplied during the recorded run: both interactive input (keyboard) and file input should be identical and also system calls should return the same result (e.g. time-of-day, system usage, ...). Since

\(^2\)A comparable technique is on-the-fly replay, as described in [15]. This system uses two parallel computers: an execution is recorded on the first one and is replayed at the same time on the second computer.
the re-execution is now deterministic, intrusive cyclic debugging tools can be used to debug the program: visualisation, data race detection, etc. is possible during replay without perturbing the original program execution.

In this survey, we start with an overview of the main sources of non-determinism in sequential, parallel and distributed programs, followed by a section on the main issues in execution replay a section on logical clocks, and two sections describing the execution replay methods that are described in literature. Finally, the paper is concluded with a conclusion, and an extensive list of references on this topic.

2 Non-determinism

In order to be able to deterministically re-execute a program, we first need to determine the non-determinism in a program execution. We start by making a distinction between different sources of non-determinism.

- **External non-determinism** means that an application returns different results for repeated executions with the same input \([10]\). This kind of non-determinism can be desired or undesired. In the cases were it is not desired, it has to be considered a bug that has to be removed from the program. In the other cases, e.g., a program that returns one solution for the eight queens problem (which has different possible solutions)\(^3\) this non-determinism has to be considered a feature of the program instead of a bug.

- **Internal non-determinism** means that repeated executions with the same input yield the same result, but the internal execution path is different. This allows to exploit different alternative executions, balancing the load, maximising the potential parallelism in the implementation, etc.

The amount of internal non-determinism depends on the level of abstraction. E.g. a program can be internally deterministic at the level of semaphore operations, but not at the level of e.g., mutexes or spinning loops used to implement the semaphores. It turns out that the amount of internal non-determinism increases with lower abstraction levels. If a program is inter-

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\(^3\)Eight queens are to be placed on a chess board in such a way that no queen threatens any other queen.
nally deterministic at the highest abstraction level, it will –by definition– be externally deterministic too.

In order to be able to re-execute an execution deterministically, one should know all the possible causes of non-deterministic behaviour. The next sections describe the different causes for sequential, parallel and distributed programs.

2.1 Sequential programs

A sequential program execution is determined by the program code and the program input. The input can be provided at program start (initial values of the memory locations) or during the execution (input from disk, keyboard, network, etc.).

Re-execution can only be deterministic if the input to the program is exactly the same as during the original execution. Replaying the input from keyboard, disk, network, and so on, is fairly straightforward to implement by logging the data during the first execution and refeeding it during the re-executions. This also applies to some system calls such as gettimeofday().

Some sources of input are harder to replay because not only the data should be refed, but this should be done at the correct moment. This applies to interrupts or signals because they cause a program transfer. This input can only be faithfully replayed by logging the moment of the interruption as well, e.g. using some kind of SIC (Software Instruction Counter) as in [3, 41].

This paper will not deal with the non-determinism caused by these types of input but will only focus on the additional non-determinism caused by the parallel or distributed nature of an application.

2.2 Parallel programs

For parallel programs one should also take into account the ‘internal’ input operations. Indeed, one could consider the memory read operations as input operations as they potentially read a value written by another process. If one executes a parallel program on a monoprocessor, we can even consider the scheduler as the cause of non-determinism instead of the many read operations. Indeed, for these machines the scheduler operations determine

\[\text{This paper assumes that the code is not changed during the execution, hence self-modifying code is not dealt with.}\]
the program execution: another execution with the same scheduling will lead to the same execution. However, this requires a scheduler that intervenes at exactly the same moment as during the first execution.

As the non-determinism caused by memory operations is an important concept for parallel programs, this type of non-determinism is called a ‘race condition’. More specifically, a race condition is defined as two unsynchronised accesses to the same shared location, and at least one access modifies it. We have to make a distinction between two types of race conditions: race conditions that are used to make a program intentionally non-deterministic: synchronisation races, and race conditions that were not intended by the programmer (data races).

We need synchronisation races to allow for competition between threads to enter a critical section, to lock a semaphore, or to implement load balancing. Removing synchronisation races makes a program completely deterministic. Therefore, in this paper, we do not consider synchronisation races a programming error, but a functional and useful characteristic of a parallel program.

Data races are not intended by the programmer, and are mostly the result of improper synchronisation. By changing the synchronisation, data races can always be removed. It is important to notice that the distinction between a data race and a synchronisation race is actually a pure matter of abstraction. At the implementation level of the synchronisation operations, a synchronisation race is caused by a genuine data race (e.g., spin locks, polling, etc.) on a synchronisation variable.

## 2.3 Distributed programs

For distributed programs, non-determinism is mainly introduced by so-called promiscuous receive operations (e.g., MPI_Recv(...,MPI_ANY_SOURCE,...) for MPI [26] programs) which can receive a message from any other process. As the source of the message is not specified, it is possible that another message is received during a re-execution. In a sense, they play the role of ‘racing’ store operations in non-deterministic programs on multiprocessors.

We assume that messages in a point-to-point communication comply with the non-overtaking property which means that successive messages sent between two nodes are always received in the order in which they were sent (this is the case for MPI and PVM [14]). Hence, since messages are assumed to be produced deterministically, receive operations in a private point-to-point
communication that specify the sender are always deterministic.

Besides the promiscuous receive operations, there is another class of instructions that can cause non-determinism in a message passing program: test operations for non-blocking message operations. These non-blocking operations return to the caller immediately: they do not wait until the message was received/delivered. In this case, test operations are used to check for the arrival of messages or to check if a send operation has finished. Non-blocking test operations are intensively used in message passing programs that use PVM or MPI, e.g., to maximally overlap communication with computation.

By the very fact that the test operations are non-blocking, they can be used in polling loops. The actual number of calls will depend on timing variations of parallel program, and is thus non-deterministic. Although many programs will not base their operation on the number of failed tests, some could do so (e.g., to implement a kind of time-out), and hence cannot be correctly replayed when not recorded.

### 3 Main issues in execution replay

For record/replay systems one will record the non-determinism at a particular abstraction level, and enforce these non-deterministic choices at the same abstraction level during the re-execution. This means that all events that happen on a higher abstraction level will be replayed faithfully, while nothing is known about the events on lower abstraction levels.

On the other hand, the level of abstraction will also determine the amount of information needed to allow for a faithful replay. The lower the abstraction level, the more information about the original execution we can obtain for the debugging session, but the more time and space we need to record the non-deterministic choices (hence more perturbation). The higher the abstraction level, the lesser we have to record, but the more uncertainty we have during debugging about the original execution. Determining the level of abstraction is hence of paramount importance and can make a record/replay system either practical or impractical. In practice, one is only interested in a re-execution that is equivalent ‘as far as the programmer is concerned’, meaning that he/she wants to have access to his/her own code, but he/she is not interested in the code of the libraries he/she is using, as long as the semantics of the library calls is preserved. An equivalent re-execution of the implementation of library routines is not needed, as the programmer cannot
observe -let alone debug- these routines anyway.

A practical execution replay system should satisfy two properties, as described in the next two sections.

### 3.1 Low overhead of the record phase

The recording overhead must be low in time \cite{11} and in space \cite{20, 22, 31}.

The \textit{time overhead} should be low to circumvent Heisenbugs and to limit the probe effect \cite{13}. A low overhead ensures that the recorded execution is more or less equivalent with a regular execution (without tracing). If the time overhead is low enough we can leave the tracing turned on all the time, even in production code. The trace can then be a standard part of a bug report. Even with a zero overhead, a program execution in record mode, is not guaranteed to be identical to the previous execution without trace mode, because no two executions are guaranteed to be identical in the presence of internal non-determinism. Leaving the tracing on all the time is the only way to guarantee a correct replay of a faulty execution.

The \textit{space overhead} for the trace files should be limited too. The first reason is that this is necessary to be able to trace long running programs. The second reason is that storing the trace requires some bandwidth, which should be shared with the target program. So, storing more means perturbing more. Therefore, the lesser is recorded, the better it is, as long as it still allows a faithful replay. As the elimination of redundant information must be done at run-time (and in real-time) the analysis should be as simple as possible, minimising program perturbation. It is clear that there is a tradeoff between the space and the time overhead. The space overhead can be limited by using the already mentioned on-the-fly replay method or by using \textit{incremental replay} techniques \cite{30}. Incremental techniques use a combination of checkpoints with execution traces. These execution traces only contain information about the execution since the last checkpoint was taken. Replay is then only possible for the part of the execution after this checkpoint, making it hard to find bugs for which there is a long time between the occurrence of the bug and the time at which the bug starts to have an effect on the execution. Furthermore, taking consistent (distributed) checkpoints is not that easy.
3.2 Faithful re-execution

The necessary condition for using cyclic debugging is that we can re-execute a program as many times as needed, and that the re-executions are in some way ‘equivalent’ with the original execution. As explained above, the abstraction level will be a critical issue. Therefore, an important question for an execution replay method is what and how much to record during the recorded execution.

There are two possible approaches to force executions of a parallel or distributed program to be equivalent to a traced execution. The first one is to force the processes to read the same values of shared variables or to receive the same messages as during the traced execution by recording the original value (content-based or data-based replay). The second one (ordering-based replay) makes sure that the interactions between the different processes occur in the same order as during the original execution. For parallel programs, the processes are forced to access the shared variables in the same order as during the traced execution, forcing the variables to undergo the modifications in the same order as during the record phase. For distributed programs, the processes are forced to receive the messages in the same order as during the original execution.

In this case, a scheme for detecting and recording the order of the operations should be used. For detecting the order in which operations are executed (or for detecting that there is no ordering and the operations are therefore in parallel), logical clocks are commonly used.

4 Logical clocks

A logical clock \[ [34, 35] C() \] should obey the so-called clock condition\(^5\)

\[ a \rightarrow b \Rightarrow C(a) < C(b) \]

This relation simple states that if \( a \) occurs causally before \( b \), the timestamp of \( a \) should be smaller than the timestamp of \( b \). As it is sufficient to order the subsequent operations on the same variable, such a clock normally calculates a new timestamp based on the old timestamp of the process and the timestamp attached to the last operation on the variable. It is clear that it

\(^5\) \( a \rightarrow b \) means that operation \( a \) ‘happened before’ operation \( b \) meaning there is some sort of causal relation between the two events.
is not sufficient to use the wall time of the operations as timestamps as the wall clock on different processes (especially in distributed machines) is not synchronised.

The simplest form of a logical clock is the scalar Lamport clock [19]: a scalar value is attached to each process. Each time a process executes an operation, the clock produces a new value: the new clock is the successor of the maximum value of the last timestamp of the process and of the timestamp attached to the last operation on the same object. This is a fairly natural way of updating the clock: the new operation happens after the last operation on the object and the last operation of the process; hence the new timestamp should be bigger than the timestamps attached to these two operations. It is clear that this type of clock satisfies the clock condition. If we have two operations $a$ and $b$ with timestamps $C(a)$ and $C(b)$ then we have three possibilities:

- $C(a) = C(b)$ meaning that there is no causal relation between these two operations.
- $C(a) < C(b)$ meaning that there could be a causal relation between these operations ($a \rightarrow b$), but we are sure that the contrary relation ($b \rightarrow a$) is not true.
- $C(a) > C(b)$ meaning the same as above with $a$ and $b$ switched.

Note that this type of clock provides enough information to get a faithful replay: if we trace the Lamport timestamps of the operations, we get a correct replay if we stall an operation $x$ with timestamp $C(x)$ until all operations $y$ with $C(y) < C(x)$ have been executed.

Vector clocks [12, 25] are used if one wants to obtain more information about an execution. A vector clock for a program with $N$ processes consists of $N$ scalar values. If process $p$ executes an operation, the new vector timestamp is calculated as follows: first, a new vector timestamp is calculated as the supremum of the last timestamps of the process and the object on which the operation is performed, and then the $p$-th element is incremented. An interesting property of a vector clock is that it not only satisfies the clock condition but also the stronger condition (meaning they are strongly consistent)

$$a \rightarrow b \Leftrightarrow C(a) < C(b)$$
This means that given two vector clocks $C(a)$ and $C(b)$, it is possible to deduce the causal relation between the two operations $a$ and $b$.

It is possible to augment the dimension of the logical clock even more, resulting in matrix clocks $[42, 40]$. Matrix clocks provide second order information to a process. It is a list of vector clocks, namely per process the last vector clock that was communicated to the current process. This information can be used to discard obsolete information in distributed systems.

5 Execution replay methods for parallel programs

In this section, an overview of the most important replay methods described in literature is given.

5.1 Content-based

In [33] a content-based replay method (name Recap) that traces the data read from every shared memory location is proposed. A trace generation of 1 MB/s on a VAX 11/780 was measured, making content-based replay impractical as the time needed to record the large amount of required information is significant, which might modify the initial execution considerably. Moreover, not only the time but also the space overhead is too large.

This replay method has however one benefit: it is possible to replay a subset of the processes (or even one process) in isolation. Nevertheless, it is argued [20] that this is no real benefit as it is difficult to examine the interactions between the different processes, hindering the task of finding the cause of a bug. Today, content-based replay is only used for tracing I/O and for tracing the result of certain system calls such as `gettimeofday()`. Indeed, this is the only viable solutions as it is not possible to ‘replay’ the operations that produced the result of these operations.

5.2 Ordering-based

As mentioned above these methods try to guarantee that the shared memory dependencies during the replay phase match the dependencies that were observed during the record phase. This will make sure that e.g. each read is executed after the same write as during the recorded execution forcing the
same data to be read. It turns out that this approach allows for a dramatical reduction in the time and space overhead in the recording phase. The tracing of the order of the memory operations can be done using hardware [4], software or hybrid probes.

5.2.1 Replaying on a monoprocessor

Replay mechanisms based on the scheduling order of the different threads can be used for monoprocessor systems. Indeed, by imposing the same scheduling order during replay, an equivalent execution is constructed [16, 39, 41]. This scheme can be extended to multiprocessor systems by also tracing the memory operations executed between two successive scheduling operations. In [7], such an implementation for Java is described. Since a typical execution of a Java program has a small number of scheduler operations (no time slicing is used and therefore scheduling is only performed at predefined points such as monitorenter calls) they succeed in producing very small trace files albeit at the cost of a large overhead (17-88%).

Such a method is of course only viable if one has complete control over the operations of the scheduler, as is only the case for proprietary systems or virtual machines such as the JVM used by Java applications.

5.2.2 Replaying on a multiprocessor

For systems where one has no control over the scheduler operations or if one wants to replay on a multiprocessor more powerful replay methods are needed. The methods described below are targeted at non-proprietary systems such as UNIX systems and require no modifications to the kernel: they all work in user space.

Instant Replay Instant Replay [20] is probably the most known replay method. The method is targeted at CREW (concurrent reader, exclusive writer) systems and attaches a version number to each shared object. Each time a read operation is performed on a certain object, the version number of the object is traced. For each write operation performed on an object, the version number of the object is incremented and the number of read operations between the last two write operations is traced. During the replay phase, each read operation stalls till the version number of the object is correct, and each write operation stalls until the same number of read
operations for a specific version (number) of the object are performed. It is clear that this method only works if the memory operations obey the CREW property: Instant Replay was proposed for a Butterfly with monitors. If a process reads or writes the object without using the monitor, the operation will not be traced and a correct replay is impossible. As such, only the synchronisation races are replayed, not the data races.

In [27], the Instant Replay method was adapted for a pSather, a parallel object-oriented programming language.

**Bacon & Goldstein** In [4], a hardware-assisted replay method was proposed. The system observes the cache traffic between the memory and the CPU’s and logs a subset of it. It is clear that this approach is highly computer dependent but it introduces no overhead at all, at the cost of extra hardware. Both the record and the replay phase use additional hardware in order to accomplish this task. A trace bandwidth of 1.17MB/s was measured for a fine-grained shared memory application on a 12-processor computer. As the tracing occurs at the lowest possible level, it is impossible to distinguish the synchronisation operations from the ‘normal’ memory operations. Therefore, this method replays both the synchronisation races and the data races.

**Netzer** In [28, 31] a replay system based on vector clocks was presented. The system attaches a vector timestamp to each process and each shared variable. Each time a process accesses a variable, both the vector timestamp of the process and the object are updated. As vector clocks are strongly consistent it is possible to detect parallel operations on the same variable by comparing the attached vector timestamps. If they turn out to be non-ordered, the operations are not properly synchronised, and as such the pair is involved in a race. By replaying the race in the same order during the replay, the race causes no harm and the replayed execution is equivalent with the recorded execution. The method succeeds in limiting the space overhead because the vector clocks automatically remove transitive ordered accesses from the trace.

Unfortunately, the method has a number of disadvantages:

- the size of a vector clock varies with the number of processes, making it difficult to deal with programs that dynamically create threads.

- a vector clock has to be attached to all shared memory locations. More-
over, the vector clocks have to compared at all shared accesses. Although the papers about the method do not contain figures about the time overhead, it is reasonable to expect a large overhead.

- data and synchronisation races are treated equally: they are both replayed. Although this sounds reasonable if one wants a correct replay, we feel that a programmer would rather remove the data race and replay the synchronisation races. After all, data races are often bugs that should be removed from the program.

**ROLT** In [23, 24] a method based on scalar Lamport timestamps was proposed. Scalar timestamps are attached to each process and each synchronisation operation. At each synchronisation operation new timestamps are calculated. For a correct replay, it is sufficient to stall each synchronisation operation until all synchronisation operations with smaller timestamps have been executed. It turns out that by logging the increments of the timestamps only, a substantial reduction of the trace files can be obtained. It is clear that by attaching scalar timestamps to all synchronisation operations, a total order is imposed on these operations although the operations in the recorded execution are only partially ordered. The additional artificial dependencies are not harmful as they do not contradict the actual execution order. They have however an impact on the replay phase as it is possible that a synchronisation operation on one synchronisation variable will have to wait for an operation on another synchronisation variable.

By using so-called fully snooped variants [3] of the Lamport clocks it is possible to calculate a total order that is consistent with the wall time of the operations, removing the synthetic dependencies from the trace (at the cost of a bigger trace file).

Compared to Netzer’s method the system has the advantage that it scales well (no vector clocks are used) and leads to small trace files. Similar to Netzer’s technique, the scalar clocks automatically calculate the transitive reduction of the order of the synchronisation operations.

The method was originally proposed for monitor operations (it was also implemented for TreadMarks [1], a distributed shared memory system [37]), making it unsuitable for programs containing data races. In [38] the replay method was extended with automatic data race detection during the replay phase. During the record phase the order of the synchronisation operations is recorded and during the first replayed execution a data race detector runs
as a watchdog. If a data race is found, the replayed execution cannot be guaranteed to be correct after the data race occurred, but this is no problem as the first data race should be removed anyway. The data race can then be removed by using cyclic debugging of the first part of the program. As soon as one is sure that an execution no longer contains data races, a regular replayed execution (without the data race detector running as a watchdog) can be used to find the non-synchronisation related errors in the particular execution.

During the data race detection phase, information about all memory operations is collected. As this would lead to a huge memory consumption, snooped matrix clocks are used in order to remove information that can be discarded (these are the memory operations that can no longer race with new operations as all new operations are causally related to the discarded ones.). The detection of the data races themselves is performed using vector clocks.

6 Methods for distributed programs

In the past, a whole deal of methods have been proposed for replaying distributed programs. Implementing such a method for distributed programs is simpler than for parallel programs as distributed programs exchange information using message operations, and these operations are typically part of a library or use a daemon (e.g. PVM and MPI). Intercepting the operations is therefore straightforward as it can be done by instrumenting function calls. This is not the case for replaying parallel programs as individual memory operations have to be traced.

Due to the large granularity one encounters in distributed applications, the time overhead is not that important. Most methods therefore primarily focus on reducing the space overhead.

Basically, two techniques are used. The first technique uses vector clocks and was proposed by Netzer in [29]. The technique basically checks whether the send operation that corresponds to a receive operation is ordered with the last receive operation of the same process. If this is true, the messages cannot be received in the wrong order, hence no race exists. If the receive and the send operations are not ordered, this is traced. As was the case for Netzer’s method for parallel programs, this method also removes transitive orders from the trace file. In [8] this method was implemented for MPI programs. The trace was augmented with information about the number of
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non-blocking test operations. Neyman et al. adapt this method for PVM-programs in [32]. Unfortunately, non-blocking test operations are not dealt with, making an exact replayed execution not always possible: any program that depends on the number of test-operations performed (e.g., for time-outs) cannot be correctly replayed.

The second technique is based on the fact that it is sufficient to trace the actual sender of messages received by promiscuous receive operations. This information can then be used during replay to force the promiscuous receive operations to wait on a messages originating from a particular sender process. Hence, a promiscuous receive operation can be made deterministic during replay by replacing it on-the-fly by a point-to-point receive operation. In [17, 6] are some examples of this method. The first method does not deal with non-blocking test operations leading to the same problems as mentioned above.

In [18] a completely different approach is taken. Instead of replaying a recorded execution on an actual computer, the replay is simulated. By altering the ‘transit time’ of messages, non-deterministic behaviour is tested. Another approach is used in [2] where only one process is partially replayed starting from a checkpoint. This method therefore offers a very limited view on an execution. The system was implemented for PVM and uses tagged messages in order to replay them.

In [36], a method for Athapascan [5], a hybrid parallel/distributed system is described. Athapascan consists of a number of nodes running on different computers that communicate using messages (on top of MPI). Each node consists of a number of POSIX threads communicating using shared memory. The proposed replay method consists two parts. The previously mentioned ROLT method is used for dealing with the non-determinism due to the shared variables. For the promiscuous receive operations, the actual sender of the messages received is recorded while the number of test operations is logged for the non-blocking test operations.

7 Conclusions

It is clear that in the last 15 year, a lot of research and work has been devoted to execution replay methods on behalf of the debugging community. However, in order to get a perfect execution replay method, one part is still missing: replaying the input. Although implementing a replay tool for input
seems straightforward—intercepting library or system calls should suffice—no tool exists at this moment, probably because this is not exactly a research field requiring much theoretical foundations.

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