FAIL-FCI: Versatile fault injection

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Abstract

One of the topics of paramount importance in the development of Grid middleware is the impact of faults, since their probability of occurrence in a Grid infrastructure and in large-scale distributed systems is actually very high. In this paper, we explore the versatility of a new tool for fault injection in distributed applications: FAIL-FCI. In particular, we show that not only are we able to fault-load existing distributed applications (as used in most current papers that address fault-tolerance issues), we are also able to inject qualitative faults, i.e. inject specific faults at very specific moments in the program code of the application under test. Finally, and although this was not the primary purpose of the tool, we are also able to inject specific patterns of workload, in order to stress test the application under test. Interestingly enough, the whole process is driven by a simple unified description language that is totally independent from the language of the application, so that no code changes or recompilation are needed on the application side.

Keywords: Fault-tolerance; Fault-injection; Stress testing; Grid middleware

1. Introduction

It is expected that Grid middleware is reliable and provides a comprehensive support for fault-tolerance mechanisms, such as failure-detection, check-pointing recovery, replication, software rejuvenation, and component-based reconfiguration, among others. One of the techniques for evaluating the effectiveness of those fault-tolerance mechanisms and the reliability level of Grid middleware is to make use of some fault-injection tools and a robustness tester to conduct some experimental assessments of the dependability metrics of the target system. In this paper, we present software that can be used both for software fault-injection and for stress testing of distributed applications, which are the basis for dependability benchmarking in Grid Computing.

Some applications (for example peer to peer applications) involve a considerable number of users, e.g. to exchange files or to execute long calculations (SeTi@Home, Decrypthon, Xtremweb, Boinc, etc.). For those applications, the appearance and disappearance of participating machines are unpredictable, very frequent, and occur while the application is run eventually. It is particularly difficult to study the functioning of large-scale distributed programs: it would be necessary to have a considerable number of computers and engineering power to execute the software in an actual situation, to measure the performances, or to detect the defects.

Testing the validity of fault-tolerant software and measuring the impact on the performance of occurring faults requires being able to control those faults. When an application is run on a cluster, it is likely that machines will run roughly at the same speed (for example a one to ten ratio on the relative speeds of the processors makes it easy to solve the consensus problem), so the considered system is actually synchronous. Afterwards, when the application is then run on a larger scale (e.g. in an Internet-like setting) where the strong synchrony hypothesis does not hold any more, crucial issues related to fault-tolerance and asynchronous settings have been overlooked.

2. Distributed fault-injection

2.1. State of the art

The issues in testing component-based distributed systems have already been described, and the methodology for testing
components and systems has already been proposed. However, testing for fault tolerance remains a challenging issue. Indeed, in available systems, the fault-recovery code is rarely executed in the test-bed, as faults rarely get triggered. As the ability of a system to perform well in the presence of faults depends on the correctness of the fault-recovery code, it is mandatory to actually test this code. Testing based on fault-injection can be used to test for fault-tolerance by injecting faults into a system under test and observing its behavior. The most obvious point is that simple tests (e.g. every few minutes or so, a randomly chosen machine crashes) should be simple to write and deploy. On the other hand, it should be possible to inject faults for very specific cases (e.g. in a particular global state of the application), even if it requires a better understanding of the tested application. Also, decoupling the fault-injection platform from the tested application is a desirable property, as different groups can concentrate on different aspects of fault-tolerance. Decoupling requires that no source code modification of the tested application should be necessary to inject faults. Finally, to properly evaluate a distributed application in the context of faults, the impact of the fault-injection platform should be kept low, even if the number of machines is high. Of course, the impact is bound to increase with the complexity of the fault scenario, e.g. when every action of every processor is likely to trigger a fault action, injecting those faults will induce an overhead that is certainly not negligible. The table below captures the major differences between the main solutions for distributed fault injection relatively to those criteria. For a more thorough survey of available tools, please refer to [9].  

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2.2. FAIL-FCI

We now describe the FAIL-FCI framework that is presented in [5]. Details of the FAIL-FCI system architecture and the FAIL language are presented in [5], while additional explanations about the software architecture are further discussed in [7].

First, FAIL (for Fault Injection Language) is a language that permits fault scenarios to be easily described. Second, FCI (for FAIL Cluster Implementation) is a distributed fault-injection platform whose input language for describing fault scenarios is FAIL. The FAIL language allows the defining of fault scenarios. A scenario describes, using a high-level abstract language, state machines which model fault occurrences. The FAIL language also describes the association between these state machines and a computer (or a group of computers) in the network.

The FCI platform is composed of several building blocks:

The FCI compiler: The fault scenarios written in FAIL are pre-compiled by the FCI compiler, which generates C++ source files and default configuration files.

The FCI library: The files generated by the FCI compiler are bundled with the FCI library into several archives, and then distributed across the network to the target machines according to the user-defined configuration files. Both the FCI compiler generated files and the FCI library files are provided as source code archives, to enable support for heterogeneous clusters.

The FCI daemon: The source files that have been distributed to the target machines are then extracted and compiled to generate specific executable files for every computer in the system. Those executables are referred to as the FCI daemons. When the experiment begins, the distributed application to be tested is executed through the FCI daemon installed on every computer, to allow its instrumentation and its handling according to the fault scenario.

3. Versatile fault injection with FAIL-FCI

In [5], the vast majority of experiments were made on a custom-made distributed program, for which both the source code and expertise were available. Moreover, our tests only dealt with the overhead of the FAIL platform, and simply showed that this overhead was, for practical purposes, negligible.

In this section, we use FAIL-FCI to inject a fault and stress test a readily available distributed application: XtremWeb [3]. The remaining of the section is organized as follows: Section 3.1 reviews the XtremWeb platform that we use for our tests. Section 3.2 describes the particular settings that we use for our experiments. Sections 3.3–3.5 describe respectively how to use FAIL-FCI for quantitative fault injection, qualitative fault injection, and stress testing. The interested reader can refer to [7] for fault injection in distributed Java implementations of Distributed Hash Tables (DHTs), and to [6] for integration with self-deploying fault tolerant MPI-based middleware.

3.1. Overview of XtremWeb

XtremWeb is a general purpose platform that can be used for high performance distributed computation. A list of tasks (or jobs) is described by the user and then distributed over the different available nodes of the system. The basic operating mode of XtremWeb is based on a participant community, e.g. it allows a High School, a University, or a Company to set up and run a Global Computing or Peer to Peer distributed system for either a dedicated application or a whole range of applications. The original XtremWeb application is written in Java, but we used here the C++ version of the software, which is expected to achieve the most efficient results. The XtremWeb tool is divided into three modules: (i) the dispatcher centralizes, organizes, and distributes the tasks, (ii) the client proposes a set of tasks to the manager, and (iii) a set of workers regularly requests a work from the manager. Like other distributed system
platforms, the XtremWeb platform uses (i) remote resources (PCs, workstations, servers) connected to the Internet, or (ii) a pool of resources (PCs, workstations, servers) inside a LAN.

3.2. Technical settings

3.2.1. Hardware settings

Each experiment was performed on two different hardware settings, to show the versatility of our fault-injection mechanism:

(1) **Cluster**: in this case, the experiments were performed on a 30-machine cluster running Linux 2.6.7. All machines have one 32 bits processor, whose frequencies vary between 1533 and 2083 MHz and whose equipped RAM is between 527 and 885 MB; they are connected through a 100 Mbps Ethernet network.

(2) **GridXplorer** [4]: in this case, the experiments were performed on 160 machines running Linux 2.6.14.3. All machines have two 64 bits 1994 MHz processors each, and 2 GB of RAM, and are connected through Gigabit Ethernet.

3.2.2. XtremWeb settings

For all the performed experiments, the XtremWeb dispatcher and client were placed on a single machine (1r17-209). The workload of the client does not really influence the dispatcher: indeed, the client and dispatcher almost run in a sequential way; the client first gives a list of jobs to the dispatcher at the beginning of the run, and the dispatcher notifies the client when the jobs have been completed and results are available. The workers are each placed on a dedicated machine in the cluster or in GridXplorer (30 and 160 such machines, respectively).

Before a particular test starts, the dispatcher is started, as well as all the workers. Then, the client is started (the staring time of the client is referred to as the **test begin time**). When the client exits (after the receipt of an acknowledgement from the dispatcher), this time is referred to as the **test end time**.

The particular application that is run with XtremWeb is **POV-Ray**, which creates three-dimensional, photo-realistic images using a rendering technique called ray-tracing. For our purpose, a task consists in calculating a particular picture using POV-Ray. This operation is requested sufficiently many times for the measures to be meaningful. When the dispatcher receives a task request from a worker, it sends all the necessary information to perform the computation of one picture.

3.3. Quantitative fault injection

We first design a probabilistic fault scenario, to quickly get a quantitative view of the fault tolerance capabilities of XtremWeb. We assume that both the dispatcher and the client are not subject to faults, (i.e. some tasks can be submitted, and some results can be returned). XtremWeb workers are run on the remaining machines, which are subject to faults. The running time is the time between when the client is started and when the results are collected. The fault model is as follows: every \( x \) seconds, each of the XtremWeb workers may crash (and cease to function) with probability \( y \). Yet, we wish to ensure that there exists a particular worker that cannot crash, in order to guarantee that the running time is always finite. The above scenario can be expressed in a surprisingly terse way using the FAIL language (with \( x = 5, y = 10\% \), and 30 workers here):

```c
spyfunc main;

Daemon ADV1 {
  node 1: before(main) -> continue, !ok(G1[1]), !go(G1), goto 2;
  node 2: }

Daemon ADV2 {
  node 1: before(main) -> stop, goto 2;
  node 2: ?ok -> continue, goto 4;
  ?go -> continue, goto 3;
  node 3: always int x = FAIL_RANDOM(1,100);
  always time_g timer = 5;
  timer & & x <= 10 -> halt, goto 4;
  timer & & x > 10 -> continue, goto 3;
  node 4: }

Computer P1 { program = "dummy"; daemon = ADV1; }
Group G1 { size = 30; program = "WorkerStatic -i 1 r17-209"; daemon = ADV2; }
```

We now informally describe the aforementioned source code. First, two automata are defined: ADV1 and ADV2; then automation ADV1 is associated to one computer P1 (that will execute the dummy code), while ADV2 is associated to 30 machines (that form the G1 group), each executing the executable file WorkerStatic with the same parameters.

ADV2 runs as follows: the daemon first waits so that the program has loaded, but before the main function is executed, the program is halted. The execution continues when the ADV1 automation sends either the ‘ok’ or the ‘go’ message. Now, the ADV1 simply sends the ‘ok’ message to a particular automation in the G1 group, and then a ‘go’ message to all automata in the G1 group. So, one automaton in the G1 group first receives an ‘ok’ message, moves to a new state (node 4), from which it simply runs the program, ignoring subsequent messages and events. So the corresponding worker process runs smoothly afterwards. In contrast, the other processes in the G1 group receive the ‘go’ message. As a result, the state is changed (node 3) so that they now receive timer events (every five seconds). When the time expires, the process under test crashes with 10% probability, while with 90% probability, the process continues its computation for another 5 s. Further details about the FAIL language can be found in [5].

We carried out this test using two values for \( x \) (5 and 10 s) and with \( y \) varying from 10% to 90% with increments of 10%. The obtained results regarding the execution time of the total set of jobs are summarized in Fig. 1. As can be seen in Fig. 1, in some of the cluster settings, the computation did not terminate (that is why there are no results for the cases every 5 s with the probability 0.8 and 0.9), due to a malfunction of the XtremWeb dispatcher (recall that this process was not purposely given a crash order by the FAIL-FCI framework). So, we also collected information about the dispatcher failure during the tests, and these results are presented in Fig. 2.

Before running the tests, one would expect that the two curves would increase, with an extra increasing gap between
them. In the Cluster setting, when there are no crashes, the time used to complete the execution of the tasks is approximately 25 s. Starting with a probability of failure of 40%, the results are as expected, but for lower probabilities, the rate of fault appearance does not significantly change the execution time. Also, when failures occur only every ten seconds, there is some kind of equilibrium (between 40% probability and 60% probability) where the execution time does not vary much. The fact that the increase is not linear reflects the fact that if more failures have occurred so far, it means that fewer failures are likely to appear (because there are a few healthy machines yet) in the future. That an equilibrium is reached is due to the fact that we ensure that at least one worker never crashes, and is thus able to complete the job if the dispatcher continues to function properly. We observe the same general phenomenon in the GridXplorer setting. But, in this setting, the phenomenon appears with a lower probability of fault appearance (between 20% probability and 30% probability), and is less marked. After this phenomenon, the two curves increase with a gap between them, like in the Cluster setting but the increase is much higher. Indeed, for the case “every 10 s”, the execution time for a probability of 30% is 2.4 times the execution time for a probability of 40%. From a 30% probability of fault appearance, the scale of the GridXplorer setting has a real impact on performance compared to the cluster setting. With a lower probability of fault appearance this impact is negligible.

When some tests did not finish, we detected in these cases that the dispatcher was still running but was not available anymore (i.e., workers could not communicate with the dispatcher to notify they completed their task). Fig. 2 shows that starting from a 70% probability for a worker to crash every five seconds, the dispatcher ends up failing in 50% of the runs. Also, from a probability of 80% for a worker to crash every five seconds, the dispatcher always fails. This failure of the dispatcher probably reveals a bug that would occur extremely rarely in a real cluster, these fault rates being pretty extreme: every 5 s, 80% of the nodes crash! A surprising observation is that in the GridXplorer setting, all tests finished despite the larger set of worker processes. So, scalability is probably not the explanation for the dispatcher malfunction in the Cluster case. Probably, the difference in machine hardware (two processors and 2 GB of RAM vs. one processor and less than 1 GB of RAM) hints that the difference is due to resource exhaustion, or perhaps a problem with thread synchronization.

3.4. Qualitative fault injection

The quantitative evaluation that was presented in Section 3.3 could also be handled, although in a more tedious and cumbersome way, through proper scripting of the distributed application. In this section, we go one step further and provide qualitative evaluation of the faults that could potentially hit the system. In more detail, we are interested here in which part of the XtremWeb workers the fault occurs. In particular, we consider the following four possible logical states for a particular XtremWeb worker:

1. job received: the XtremWeb worker has received a job to perform from the XtremWeb dispatcher,
2. after compute: the XtremWeb worker has finished performing its task,
3. job finished: the XtremWeb worker has notified the XtremWeb dispatcher that it has completed its job,
(4) **job completed**: the XtremWeb worker has sent the XtremWeb dispatcher the results of the completed task.

Our goal in this series of tests is to fix the number of workers (30 and 160, respectively) and the crash probability (40%), but a worker may only fail at precise points in its program code: the points that correspond to entering the four states mentioned above. The corresponding FAIL program (i.e., fault scenario) is as follows (considering that faults would only occur when the worker is in the state **job completed**):

```c
spyfunc main; spyfunc Protocol::DataSaved;
spyfunc Protocol::release;

Daemon ADV1 {
  node 1: before(main) -> continue, !ok(G1[1]), !go(G1),
         goto 2;
  node 2: 
}

Daemon ADV2 {
  node 1: before(main) -> stop, goto 2;
  node 2: ?ok -> continue, goto 5;
  ?go -> continue, goto 3;
  node 3: always int x = FAIL_RANDOM(1,100);
         before(Protocol::DataSaved) && x <= 40 -> continue,
         goto 4;
         before(Protocol::DataSaved) && x > 40 -> continue,
         goto 3;
  node 4: before(Protocol::release) -> stop, goto 5;
  node 5: 
}
```

As in Section 3.3, there are two automata ADV1 and ADV2 that are dispatched in the same way as before. The same trick to get at least one working worker is also used (using the ‘ok’ and ‘go’ messages). The key difference is the use of breakpoints to get back control over the processes when a particular function is reached. In this scenario, the methods **DataSaved** and **release** of the class **Protocol** are watched. The state **job completed** is reached after the call to the method **DataSaved** has completed, and just before the call of the method **release**. Note that the **release** method is called often and in various contexts in the XtremWeb worker code, but only corresponds to the **job completed** state after the **DataSaved** method has been executed.

The obtained results are summarized in Fig. 3. In this figure, the category **without fault** refers to the test without injecting faults (for comparison purpose). For each of the four aforementioned possible states of the workers, two kinds of faults are considered:

1. suspending the process (using **stop** in the **FAIL** language) to simulate an overloaded machine,
2. crashing the process (using **halt** in the **FAIL** language).

We did not collect information about possible dispatcher failures, since no crashes were observed in both settings (this was expected in the Cluster setting, because the probability of crashes was 40%).

It was expected that injecting **stop** faults would induce a worse performance than injecting **halt** faults (because in the first case, the other end of the TCP connexion, i.e. the dispatcher, is not notified by the network layer that something bad happened, while in the second case, it usually is). This was confirmed by the results we obtained on the cluster and on GridExplorer. We also expected that the later the injection (but yet before the results are sent to the dispatcher), the more time it would take to complete the computation. However, and surprisingly, if the workers crash before even starting a computation, the performance is worse than if it crashes after the computation. This behavior, which appears in the two settings, is probably due to a misconception in the XtremWeb dispatcher, which does not expect failures just after the job was sent (at that time, it is probably not watching the TCP connection with the worker, while it is when the job is near to completion). We also remark that if a worker crashes after a job is completed (and the worker notified the controller that the results are available), then the performance is almost the same as if no faults were injected. Thus, the scale is not really an issue when dealing with qualitative faults, as we are interested in relative performance issues.

### 3.5. Stress testing

Sections 3.3 and 3.4 showed how **FAIL-FCI** can be used to obtain failure resilience capabilities of distributed applications using a unified approach for both quantitative and qualitative analysis. We now show that the same tool can be set up to handle stress testing as well. We understand “stress testing” here in a very specific sense: We wish to evaluate the dispatcher’s performance when the number of workers that simultaneously arrive in the system increases. The goal is to measure the overhead for the dispatcher to manage workers. When the number of available workers increases, the computational power also increases, but the dispatcher may have to manage more simultaneous connections, and the global performance may not be necessarily better. For this purpose, we use a slightly different scenario. The set of tasks is the same as before, and the XtremWeb client and dispatcher are still on the same machine. The **test begin time** is the time when both the XtremWeb client and dispatcher are up and running, waiting for workers to perform the tasks. Then, a particular XtremWeb worker is launched into action with probability \( y \) every \( x \) seconds. When the client exits (after having received the acknowledgement from the dispatcher), the current time is taken as the **test end time**.

The corresponding scenario written using the **FAIL** language is as follows (considering that \( x = 1 \) and \( y = 10\% \)):

```c
spyfunc main;

Daemon ADV1 {
  node 1: before(main) -> continue, !go(G1), goto 2;
  node 2: 
}

Daemon ADV2 {
  node 1: before(main) -> stop, goto 2;
  node 2: ?go -> stop, goto 3;
  node 3: always int x = FAIL_RANDOM(1,100);
         always time && time = 1;
         time && x <= 10 -> continue, goto 4;
         time && x > 10 -> stop, goto 3;
  node 4: 
}
We performed tests by varying $x$ from 1 to 9 s (with increments of 2 s), and varying $y$ from 10% to 100% in the Cluster setting, and varying $x$ from 1 to 2 s (with increments of 2 s), and varying $y$ from 10% to 100% in the GridExplorer setting. The obtained results regarding the global execution time are summarized in Fig. 4. We did not collect information about dispatcher failure, since none appeared.

It was expected that the shape of the curves would have a "U" form for at least the test with $x = 1$ (getting workers to the job every second): if few workers arrive at the same time, the performance is low; if several workers arrive at the same time, this is manageable by the dispatcher and the performance is good; if many workers arrive at the same time, the dispatcher would be more overloaded and the overall performance would be worse than with fewer workers. In the Cluster setting, all curves are decreasing, which means that the more workers you get, the faster is the completion of the computation. It also means that the C++ version of XtremWeb can handle 30 new workers arriving at the same time with no problems (this is the case where $y = 100\%$). On the contrary, the GridExplorer setting shows that the more workers you put simultaneously in the system, the longer it takes to complete the task. So, with 160 workers, many of which arrive simultaneously, the XtremWeb dispatcher is clearly stressed but degrades gracefully.

**4. Concluding remarks**

We proposed a unified approach for fault injection and stress testing distributed applications. Fault injection can be made using a quantitative approach (as in most related studies) as well as the more original qualitative approach, where precise faults are inserted at precise logical states of the application under test. Although the set of possible fault injections is extremely large, the language that describes the faults scenario is high level and independent from the language used in the application. This enables decoupling between the application programmers and the test specifiers, so that their expertise is used in the proper domain.

The results we obtained on XtremWeb show that it suffers from a problem of resource management, in the dispatcher component, when workers crash, which often leads to failure of the whole computation. Another possible improvement would be to properly take care of worker failure when faults appear just after a job reception, the resulting performance being
worse than in the case where the fault appears right after the computation at the worker (although in this last case the worker has lost time in the computation).

As a proof of concept, we also showed that the same specification language and fault injection tool could be used as a stress test platform as well. The preliminary tests we performed actually raised a number of interesting open questions. The main one relates to the use of FAIL-FCI at a larger scale (for purpose of stress testing), i.e. 1000-10,000 machines. We are currently investigating using our fault injector in larger systems, typically by using emulation schemes, within the GridExplorer project platform. Further studies are needed to see the effect of correlated faults injections (such as those occurring when a virus is spread throughout the network).

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References


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