

Dynamic Symbolic Execution for Testing Distributed Objects [★]

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Abstract. This paper extends dynamic symbolic execution to distributed and concurrent systems. Dynamic symbolic execution can be used in software testing to systematically identify equivalence classes of input values and has been shown to scale well to large systems. Although mainly applied to sequential programs, this scalability makes it interesting to consider the technique in the distributed and concurrent setting as well. In order to extend the technique to concurrent systems, it is necessary to obtain sufficient control over the scheduling of concurrent activities to avoid race conditions. Creol, a modeling language for distributed concurrent objects, solves this problem by abstracting from a particular scheduling policy but explicitly defining scheduling points. This provides sufficient control to apply the technique of dynamic symbolic execution for model based testing of interleaved processes. The technique has been formalized in rewriting logic, executes in Maude, and applied to non-trivial examples, including an industrial case study.

1 Introduction

Distributed and concurrent systems, e.g. web services, are becoming increasingly important for long-running infrastructure and applications. They typically consist of loosely coupled components which communicate asynchronously, potentially running on different hardware systems. For critical distributed systems, the use of formal methods, both for design and verification, remains a challenge. In the general case, the complexity of such systems makes full verification seem impossible, even for medium sized examples. In this paper we consider model-based testing of distributed concurrent systems, where we use an object oriented, distributed model as specification.

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We present a tool which identifies adequate test cases from such a formal model. In order to test the different communication patterns, we focus on architectural models which reflect the distributed nature of the systems under test. Hence, the models themselves are complex in the sense that they have to capture distribution, concurrency, and asynchronous communication. The challenge is to find a test generation technique that scales to the combinatorial explosion in the number of possible runs in such models. A promising technique that seems to scale well to large systems is *dynamic symbolic execution* [2, 8, 18, 19]. The idea is to calculate a symbolic execution in parallel with the concrete test run of a given formal model. The result is a set of conditions over symbolic input values representing the path of the last run. The conjunction of these conditions form the equivalence class of inputs that could take the same path.

The problem is that dynamic symbolic execution cannot deal with common concurrency models as present in today's programming languages. The reason is that dynamic symbolic execution does not work in the context of arbitrary non-deterministic interleavings of executions. Hence, its main application so far has been limited to single-threaded (sequential) programs and to client-server applications with simple serialized communication flows. In this work we overcome this limitation by choosing a modeling language that provides the appropriate level of concurrency control: Creol [10].

Creol is an executable object oriented modeling language whose execution model was designed to assist in the development of distributed systems. An object in Creol describes an execution unit that executes a dynamic number of processes, a single process at a time. Features like asynchronous method calls and conditional release points allow to model complex interactions between distributed components or objects.

We have implemented the dynamic symbolic execution technique in Maude [4], which is the execution platform of Creol, allowing us to perform the symbolic run dynamically while the concrete run is executed. The tool computes the equivalence classes of test inputs covering the paths already taken, allowing the tester to systematically find new test stimuli for non-covered parts. The generated test cases are used to check the conformance of implementations of the distributed systems with their Creol models as presented in previous work in [1]. The presented technique forms part of a new design process for distributed systems that has been developed in the EU FP6 CREDO project. The feasibility of the approach has been shown by application to the ASK system, an industrial distributed agent-based information system.

To summarize, the *contributions* of this work are as follows:

- This is the first time dynamic symbolic execution is applied to distributed systems involving asynchronous method calls and non-deterministic scheduling of interleaved processes.
- The technique has been formalized in terms of rewriting logic and implemented in the Maude rewriting system.
- It has been applied to an industrial case study.

In the remainder of this section we give an overview of related work, followed by a short introduction to dynamic symbolic execution in the next section and an introduction to Creol in Section 3. Dynamic symbolic execution is extended to distributed systems in Section 4 and applied to testing in Section 5, before showing examples in Section 6. Finally, in Section 7 we draw our conclusions.

1.1 Related Work

Symbolic execution is a widely used program analysis technique that represents the values of variables as symbolic expressions instead of concrete data. An execution of a program is performed by manipulating those expressions instead of computing concrete values. Application of symbolic execution to testing was already proposed in 1976 by King [12], who shows symbolic execution for a simple sequential language and presents an interactive tool EFFIGY to traverse the execution tree.

Much more recently, symbolic execution has been used for various applications in the area of testing. Khurshid et al. [11] perform source to source transformation on Java programs to allow explicit state model checkers like the Java PathFinder [21] to exploit the succinct representation of the state space by symbolic representation. They generate test cases by checking the reachability of a testing criterion. Analysis of the counter example gives the input for test cases similar to [22, 7, 9]. In [23], Xie et al. introduce SYMSTRA, a tool that uses symbolic execution to explore different sequences of method calls in order to generate unit tests for object oriented systems. These applications use symbolic execution mainly to compress the representation of the state space while performing an exhaustive search. However, there are limits to the feasibility of executing complex concurrent systems purely symbolically, due to the sheer number of possible execution paths induced by non-determinism.

There are basically two possibilities to make the process feasible for large systems: (1) reducing the amount of information which needs to be tracked and (2) reducing the number of paths to search. An example for the first kind are static analysis tools like ARCHER from Engler et al. [24], which very successfully concentrate on certain properties of interest for the analysis (memory and array access). To derive input values that drive a run to certain areas in the program, however, we want to consider all information available. We therefore reduce the number of paths that are searched at the same time to make symbolic execution feasible. The latter technique is called *dynamic symbolic execution* (DSE).

To our knowledge, the first to use symbolic execution on single runs were Boyer et al. in 1975 [2] who developed the interactive tool SELECT that computes input values for a run selected by the user. One of the first automated DSE tools for testing was DART (Directed Automated Random Testing) from Godefroid et al. [8]. DART automatically extracts a program's interface and generates a test driver to perform random testing. While DART only evaluates integer variables, the CUTE and jCUTE tools from Sen et al. [18] extend this approach to include pointers and generate dynamic data structures. Several extensions to these approaches exist, among the most notable the PEX tool from

Tillmann et al. [19] for computing test cases using *parameterized unit tests* for single-threaded .NET programs.

We extend this approaches to a model-based testing method that targets distributed and concurrent systems and deals with interacting processes and asynchronous communication between components. Model-based testing uses models of the system under test (SUT) to derive test cases. Evaluations, e.g., from Pretschner et al. [17], have shown their usefulness in software development. Tools for reactive systems, like TorX from Tretmans et al. [20], observe the inputs of the SUT and perform on-the-fly testing by generating new inputs for the SUT according to the model and a test purpose. In contrast to this applications, distributed systems are only loosely coupled, close synchronization between SUT and tester is not useful as we discussed in more detail in previous work [1]. In our setting, the specification is given as Creol model. Creole is a modeling language whose semantics is defined in rewrite logic, which is executable in Maude. The definition of the language in rewrite logic therefore directly gives an interpreter, an approach that also was used by Chen et al. [3] for their framework for “rapid prototyping” of new languages. In this work, we extend the semantic rules to perform DSE on Creol models in order to find test cases with optimal coverage of this specification. We compute test suites to check the conformance between an implementation and the specification. Our previous work [1] also shows how to use a Creol model as an oracle for a test run on the implementation.

Recent work from Kirner [13] gives criteria to ensure that a coverage metric on the model also holds on the actual implementation. In this work we assume that these criteria hold. There are a number of techniques that help in testing of concurrent systems by either controlling the scheduling to make the test results more deterministic [14, 16] or by repeating test cases multiple times with a different (randomized) scheduling to gain a good coverage of the code [6]. These methods are complementary to the approach shown here as they handle the actual test execution rather than the computation of test cases and should be combined with the test case generation shown in this paper for optimal results.

2 Dynamic Symbolic Execution for Testing

This section gives a brief introduction to dynamic symbolic execution (DSE) and its application to conventional test case generation, before we proceed with extensions for distributed and concurrent systems. Conventional symbolic execution uses symbols to represent arbitrary values during execution. When encountering a conditional branch statement, the run is forked. This results in a tree covering all paths in the program. Decisions on branch statements are recorded, resulting in a set of conditions over the symbolic values that have to evaluate to *true* for a path to be taken. We call the conjunction of these conditions the *path condition*; it represents an equivalence class of concrete input values that could have taken the same path. In contrast, *dynamic* symbolic execution calculates the symbolic execution *in parallel* with a concrete run that is actually taken, avoiding the

usual problem of eliminating infeasible paths and maintaining the call stack of the whole run tree.

We use DSE to compute test cases on the model that are then used on the actual implementation. The inputs of the model are treated as symbolic values and a path condition describing the equivalence class of inputs that can perform the same run is computed. Concrete input values from *outside* this equivalence class are selected to force new execution paths, and thereby new test cases. Consider the following piece of code from an agent system calculating the number of threads needed to handle job requests (taken from Figure 5).

```

1   amountToCreate:= tasks - idlethreads + ... ;
2   if (amountToCreate > (maxthreads - threads)) then
3       amountToCreate:=maxthreads - threads;
4   end;
5   if (amountToCreate > 0) then ... end;

```

Testers usually analyze the control flow in order to achieve a certain coverage. For example, a run evaluating both conditions above to **true** is sufficient to ensure *statement coverage*; *branch coverage* needs at least two cases and *path coverage* all four combinations. Dynamic symbolic computation gives a condition for each conditional statement in terms of symbolic input values. For better readability, we mark the symbolic values of an input parameter by appending S to the parameter's variable name. Let `threads`, `idlethreads`, and `tasks` denote the input parameters for testing, and `maxthreads` being a constant with a concrete value. Assume that we have a first concrete run in which both conditions evaluate to **true**. This single run already fulfills statement coverage. DSE gives us following path condition that has to be fulfilled to obtain the run (for constant `maxthreads = 10`)

$$(tasks_S - idlethreads_S) > (10 - threads_S) \\ \wedge (maxthreads_S - threads_S) > 0$$

In this example, DSE replaces the variable `amountToCreate` by its symbolic value `maxthreads_S - threads_S`. To create a new test case that follows a different path, one or more of the sub-conditions are negated and inputs that fulfill the new condition are selected. If the path condition is not satisfiable, the corresponding path is infeasible. In this case, we continue negating different sub-conditions until no more valid inputs are found. For example, inputs satisfying

$$(tasks_S - idlethreads_S) \leq (10 - threads_S) \\ \wedge (maxthreads_S - threads_S) > 0$$

will avoid the first **then**-branch, resulting in a different execution path.

The strategy how to select sub-conditions to negate determines the kind of coverage metric obtained. Note that the fraction of the program that can be covered depends on the program, the used coverage metric, and the symbolic values used. For example, the presence of unreachable code obviously makes full statement coverage impossible. The concrete test values from symbolic input vectors can be found by, e.g., using a constraint solver.

3 The Modeling Language Creol

Creol is a high-level executable modeling language targeting distributed systems in which concurrent objects communicate asynchronously [10]. The language decouples communication from synchronization. Furthermore, it allows local scheduling to be left underspecified but controlled through explicitly declared process release points. The language has a formal semantics defined in rewriting logic [15] and executes on the Maude platform [4]. In the remainder of this section, we present Creol and point out its essential features for DSE.

A concurrent object in Creol executes a number of processes that have access to its local state. Each process corresponds to the activation of one of the object's methods; a special method `run` is automatically activated at object creation time, if present, and captures the object's active behavior. Objects execute concurrently: each object has a processor dedicated to executing the processes of that object, so processes in different objects execute in parallel. In contrast to, e.g., Java, each Creol object strictly encapsulates its state; i.e., external manipulation of the object state happens via calls to the object's methods only.

Only one process can be active in an object at a time; the other processes in the object are *suspended*. We distinguish between *blocking* a process and *releasing* a process. Blocking causes the execution of the process to stop, but does not let a suspended process resume. Releasing a process suspends the execution of that process and lets another (suspended) process resume. Thus, if a process is blocked there is no execution in the object, whereas if a process is released another process in the object may execute. The execution of several processes within an object can be combined using *release points* within method bodies. At a release point, the active process may be released and *some* suspended process resumes. This way, (non-terminating) active and reactive behavior are easily combined within a concurrent object in Creol.

Communication in Creol is based on method calls. These are a priori asynchronous; method replies are assigned to labels (also called *future variables*, see [5]). There is no synchronization associated with *calling* a method. *Reading a reply* from a label, however, is a blocking operation and allows the calling object to synchronize with the callee. A method call that is directly followed by a read operation models a synchronous call. Thus, the calling process may decide at runtime whether to call a method synchronously or asynchronously. The local scheduling of processes inside an object is given by conditions associated with release points. These conditions may depend on the value of the local state, allowing cooperative scheduling between the processes within an object, but may also depend on the object's communication with other objects in the environment. Guards on release points include synchronization operations on labels, so the local scheduling can depend on both the object's state and the arrival of replies to asynchronous method calls.

To sum up: only one process is executing on each object's local state at a time, and interleaving of processes is flexibly controlled via (guarded) release points. Together with the fact that objects communicate exclusively via messages (strict

$T ::= C \mid \mathbf{Bool} \mid \mathbf{Void}$ $\quad \mid \mathbf{Int} \mid \mathbf{String} \mid \dots$ $v ::= f \mid x$ $b ::= \mathbf{true} \mid \mathbf{false} \mid v$ $g ::= b \mid v? \mid g \wedge g$	$L ::= \mathbf{class} \ C(\bar{v}) \ \mathbf{begin} \ \overline{\mathbf{var} \ f : \bar{T}; \bar{M}} \ \mathbf{end}$ $M ::= \mathbf{op} \ m(\mathbf{in} \ \bar{x} : \bar{T} \ \mathbf{out} \ \bar{x} : \bar{T}) \ == \ \mathbf{var} \ x : \bar{T}; \bar{s} \ \mathbf{end}$ $e ::= v \mid \mathbf{new} \ C(\bar{v}) \mid \mathbf{null} \mid \mathbf{this} \mid v + v \mid \dots$ $s ::= l!.m(\bar{e}) \mid !e.m(\bar{e}) \mid l?(v) \mid e.m(\bar{e}; \bar{v}) \mid \mathbf{await} \ g$ $\quad \mid v := e \mid \mathbf{skip} \mid \mathbf{release} \mid \mathbf{await} \ e.m(\bar{e}; \bar{v})$ $\quad \mid \mathbf{while} \ g \ \mathbf{do} \ \bar{s} \ \mathbf{end} \mid \mathbf{if} \ g \ \mathbf{then} \ \bar{s} \ \mathbf{end}$
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Fig. 1. Language syntax of a subset of Creol.

encapsulation), this gives us the concurrency control necessary for extending DSE to the distributed paradigm.

Syntax. The language syntax of the subset of Creol used in this paper is presented in a Java-like style in Figure 1. In this overview, we omit some features of Creol, including interfaces, inheritance, non-deterministic choice and many built-in data types and their operations. For a full overview of Creol, see for example [10]. In the language subset used in the examples of this paper, classes L are of type C with a set of methods \bar{M} . *Expressions* e over variables v (either fields f or local variables x) are standard. *Statements* s are standard apart from the asynchronous method call $l!.m(\bar{e})$ where the label l points to a reference to the reply, the (blocking) read operation $l?(\bar{v})$, and release points **await** g and **release**. *Guards* g are conjunctions of Boolean expressions b and synchronization operations $l?$ on labels l . When the guard in an **await** statement evaluates to *false*, the statement is *disabled* and becomes a **release**, otherwise it is *enabled* and becomes a **skip**. A **release** statement suspends the active process and another suspended process may be rescheduled. The *guarded call* **await** $e.m(\bar{e}; \bar{v})$ is a typical pattern which suspends the active process until the reply to the call has arrived and abbreviates $l!.m(\bar{e}); \mathbf{await} \ l?; l?(\bar{v})$.

3.1 Representation of a Run

A run of a Creol system captures the parallel execution of processes in different concurrent objects. Such a run may be perceived as a sequence of execution steps where each step contains a set of local transitions on a subset of the system's objects. However, only one process may be active at a time in each object and different objects operate on disjoint data. Therefore, the transitions in each execution step may be performed in a truly concurrent manner or in any sequential order, so long as all transitions in one step are completed before the next execution step commences. For the purposes of dynamic symbolic execution the run is represented as a sequence of statements which manipulate the state variables, together with the conditions which determine the control flow, as follows.

The representation of an assignment $\bar{v} := \bar{e}$ is straightforward: Because fields and local variables in different processes can have the same name and statements from different objects are interleaved, the variable names are expanded to a unique identifier by adding the object id for fields and the call label for local variables. This expansion is done transparently for all variables and we will omit the variable scope in the following.

An asynchronous method call in the run is reflected in four execution steps (the label value l uniquely identifies the steps that belong to the same method call): $o_1 \xrightarrow{l} o_2.m(\bar{e})$ represents the *call* of method m in object o_2 from object o_1 with arguments \bar{e} ; $o_1 \xrightarrow{l} o_2.m(\bar{v})$ represents when the called objects starts execution, where \bar{v} are the local names of the parameters for m ; $o_1 \xrightarrow{l} o_2.m(\bar{e})$ represents the emission of the return values from the method execution; and $o_1 \xleftarrow{l} o_2.m(\bar{v})$ represents the corresponding reception of the values. These four events fully describe method calling in Creol. In this execution model the events reflecting a specific method call always appear in the same order, but they can be interleaved with other statements.

Object creation, **new** $C(\bar{v})$, is similar to a method call. The actual object creation is reduced to generating a new identifier for the object and a call to the object's **init** and **run** methods, which create the sequences as described above.

Conditional statements in Creol do not change the values of the variables and therefore can be treated as **skip** in DSE. For the sake of computing the input values, however, the condition of the taken branch is recorded as $\langle g \rangle$ (E.g., if the concrete execution selects the *then* branch of an statement **if** g , the condition $\langle g \rangle$ is recorded. If the *else* branch is selected, then the *negated* condition $\langle \neg g \rangle$ is recorded). Remark that statements **await** g requires careful treatment: if it evaluates to *false*, no code is executed. To reflect the information that the interpreter failed to execute a process because the condition g of the **await** statement evaluated to *false*, the negated condition $\langle \neg g \rangle$ is recorded and the interpreter proceeds by selecting another process.

4 Dynamic Symbolic Execution of Distributed Objects

This section presents the rules to actually compute the symbolic values for a given run. The formulas given in this section very closely resemble the rewrite rules of Creol's simulation environment [10], defined in rewriting logic [15] and implemented in Maude [4]. A rewrite rule $t \Longrightarrow t'$ may be interpreted as a *local transition rule* allowing an instance of the pattern t in the configuration of the rewrite system to evolve into the corresponding instance of the pattern t' (where t and t' denote *states* of the model). When auxiliary functions that do not change the state are needed in the semantics, these are defined in equational logic, and are evaluated in between the state transitions [15]. The rules are presented here in a slightly simplified manner to improve readability.

Denote by \bar{s} the representation of program statements. Let $\sigma = \langle v_1 \triangleright e_1, v_2 \triangleright e_2, \dots, v_n \triangleright e_n \rangle = \langle \bar{v} \triangleright \bar{e} \rangle$ be a map which records *key-value* entries $v \triangleright e$, where a variable v is bound to a symbolic value e . The value assigned to key v is accessed by $v\sigma$. For an expression e and a map σ , define a parallel substitution operator $e\sigma$ which replaces all occurrences of every variable v in e with the expression $v\sigma$ (if v is in the domain of σ). For simplicity, let $\bar{e}\sigma$ denote the application of the parallel substitution to every expression in the list \bar{e} . Furthermore, let the operator $\sigma_1 \uplus \sigma_2$ combine two maps σ_1 and σ_2 such that, when entries with the same

$$\begin{aligned}
\bar{v} := \bar{e}; \bar{s}[\Theta, \sigma, \mathcal{C}] &\Longrightarrow \bar{s}[\Theta, \sigma \uplus \langle \bar{v} \triangleright (\bar{e}\sigma) \rangle, \mathcal{C}] && \text{(ASSIGN)} \\
o_1 \xrightarrow{l} o_2.m(\bar{e}); \bar{s}[\Theta, \sigma, \mathcal{C}] &\Longrightarrow \bar{s}[\Theta \uplus \langle l \triangleright \bar{e}\sigma \rangle, \sigma, \mathcal{C}] && \text{(CALL)} \\
o_1 \xrightarrow{l} o_2.m(\bar{v}); \bar{s}[\Theta, \sigma, \mathcal{C}] &\Longrightarrow \bar{s}[\Theta, \sigma \uplus \langle \bar{v} \triangleright l\Theta \rangle, \mathcal{C}] && \text{(BIND)} \\
\langle g \rangle; \bar{s}[\Theta, \sigma, \mathcal{C}] &\Longrightarrow \bar{s}[\Theta, \sigma, \mathcal{C} \hat{\ } \langle g\sigma \rangle] && \text{(COND)}
\end{aligned}$$

Fig. 2. Rewrite rules for symbolic execution of Creol statements.

key exist in both maps, the entry in σ_2 is taken. These operators are defined as equations in rewriting logic and are evaluated in between the rewrite steps. In the symbolic state σ , all expanded variable names are bound to symbolic expressions. However, operations for method calls do not change the value of the symbolic state, but generate or receive *messages* that are used to communicate actual parameter values between the calling and receiving objects. Similar to the expressions bound to variables in the symbolic state σ , the symbolic representations of these actual parameters are bound in a map Θ to the actual and unique label value l provided for each method call by Creol’s operational semantics. Finally, the conditions of control statements along an execution path are collected in a list \mathcal{C} ; the concatenation of a condition c to \mathcal{C} is denoted by $\mathcal{C} \hat{\ } c$.

The *configurations* of the rewrite system for dynamic symbolic execution are given by $\bar{s}[\Theta, \sigma, \mathcal{C}]$, where \bar{s} is a run represented as a sequence of statements that still have to be executed, Θ and σ are the maps for messages and symbolic variable assignments as described above, and \mathcal{C} is the list of conditions. Recall that the run \bar{s} (as described in Section 3.1) is in fact generated on the fly by the concrete rewrite system for Creol executed in parallel with the dynamic symbolic execution. Thus, the *rules* of the rewrite system have the form

$$\bar{s}[\Theta, \sigma, \mathcal{C}] \Longrightarrow \bar{s}'[\Theta', \sigma', \mathcal{C}']$$

The primed versions are updated results from the execution rule. The rules are given in Figure 2 and explained below.

Rule ASSIGN defines the variable updates that are performed for an assignment. All variables in the right hand side are replaced by their current values in σ , which is then updated by the new expressions. Note that we do not handle variable declarations, but work in the runtime-environment. We expect that a type check already happened during compile time and insert variables into σ the first time they appear. A method call as defined by Rule CALL emits a message that records the expressions that are passed to the method. Because of the asynchronous behavior of Creol, the call might be received at a later point in the run (or not at all if the execution terminates before the method was selected for execution) by Rule BIND, which handles the binding of a call to a new process and assigns the symbolic representation of the actual parameter values to the local variables in the new process. The emission and reception of return values are handled similarly to call statements and call reception.

Object creation is represented as a call to the constructor method `init` of the newly created object. In this case there is no explicit label for the call statement,

so the object identifier is used to identify the messages to call the `init` and `run` methods, which are associated to the new statement. For conditionals, the local variables in the condition are replaced by their symbolic values (Rule COND). This process is analogous for the different kinds of conditional statements (**if**, **while**, **await**). The statement itself acts as a **skip** statement; it changes no variables and does not produce or consume messages. The resulting expression $g\sigma$ directly characterizes the equivalence class of input values that reach and fulfill the condition. The conjunction of all conditions found during symbolic evaluation give the set of input values that can perform that run. The tool records the condition that evaluated to *true* during runtime. Therefore, if the **else** branch of an **if** statement is entered or a disabled **await** statement with g approached, the recorded condition will be $\neg g$.

5 Testing Distributed Systems

Approaches to test case generation for structural coverage intend to find test a set that performs runs in the system for a specific coverage criterion. Two runs that cover the same parts of a system are considered equivalent. A good test set should maximize the coverage, while minimizing the number of equivalent runs in order to avoid superfluous efforts in executing the tests.

The execution of a distributed system is not fully controllable through its interface. One and the same test case can lead to arbitrarily different runs on the system under test (SUT). In practice, tools like ConTest [6] are used to execute single test cases multiple times on the SUT with different schedulings. For the model, on the other hand, it is straightforward to introduce additional variables to resolve the nondeterminism for the sake of examining all possible paths to build the optimal set of test cases. These techniques are complementary to the computation shown in this paper and should be applied additionally.

It is the responsibility of a testing engineer to write test objects (analogous to unit tests) that set up the system and perform interactions that will drive an interesting execution of the system. Presupposing this test scenario, we enhance the coverage by introducing symbolic values t_S in the test object and compute new values such that new, non-equivalent runs are performed.

Constructing the Test Set. Dynamic symbolic execution on a run gives the set of conditions that are combined to the path condition $\mathcal{C} = \bigwedge_{1 \leq i \leq n} c_i$ (for n conditions), characterizing exactly the equivalence class of t_S that can repeat the same execution path. Only one test case that fulfills \mathcal{C} is selected. A new test case is then chosen to specifically avoid that a particular branch is taken by violating the respective c_i . To maximize decision coverage (DC), for instance, test cases have to be created such that for each of the conditions c_i , there is also a test case that violates this condition. The process of generating new test cases ends after all combinations required for the coverage criteria are explored.

In the case of concurrent distributed systems, however, we frequently deal with scenarios in which the naive approach does not terminate. Most importantly, distributed systems usually contain active objects that do not terminate

and thus creates an infinite run. In this case, execution on the model has to be stopped after exceeding some threshold (ideally after detecting a loop). The computation of the condition can be performed as before and will prohibit the same partial run in future computations. Creol also supports infinite datatypes. Therefore, for a code sample like **while** ($i > 0$) **do** $i := i - 1$ **end**, there is a finite run for each i , but there are infinitely many of them. To make sure that the approach terminates, an artificial limiting condition has to be introduced, e.g., by creating an equivalence class for all i greater than a constant k .

Running a Test Case. A test case as generated in this paper is used to test implementations of distributed systems by checking if the implementation under test complies to the model as described in previous work [1]. The test execution approach of that paper handles the difficulties of testing a distributed system by defining a set of actions and events that are used to control the implementation as well as the model, and to monitor the behavior of the implementation. So far the execution has not been monitored online, rather a log is generated that has been verified by using the model. A run of the implementation is considered successful if the model is able to reproduce the run.

The model is a direct specification of the implementation, and both systems share their internal control structure. Test cases optimized for structural coverage in the model will therefore also improve the structural coverage in the implementation.

6 Examples

This section shows the feasibility of the approach by means of two examples: The *peer to peer* example presents the exploration of existing test cases with respect to coverage, during which an important special case was discovered. The second example demonstrates how to derive new test cases on example of the *ASK system*, an industrial case study.

The dynamic symbolic interpreter allows to identify variables that are treated as normal variables for the concrete run, and as a symbolic value for the dynamic symbolic execution. These variables are identified by a special naming scheme, here denoted by the subscript \mathcal{S} . This enables the flexible monitoring of symbolic values of variables at any arbitrary level in the code.

6.1 Peer to Peer

A peer to peer system connects several coequal components (peers) with the aim to share data between them. Each peer works both as client and as server holding local files. A client can search the network to find the location of a file, connect to the respective server and download the document. Communication between the components is established via channels. We use a sophisticated model describing such a system, which stems from the CREDO project to demonstrate various techniques for modeling distributed systems. It consists of 23 classes (not shown in this paper due to lack of space) and already comes with a small set of test

```

1 class Test (cl :Client, b :Peer)
2 begin
3   var reply :Data
4   op run ==
5     await
6       cl.search("f1";reply)
7
8
9 end

```

Fig. 3. Predefined test case

```

1 class TestDSE (cl :Client, b :Peer)
2 begin
3   var reply :Data
4   op run ==
5     var reqkeyS :Data;
6     reqkeyS := "f1" ;
7     await
8       cl.search(reqkeyS;reply)
9 end

```

Fig. 4. Test case for DSE

cases that model a net consisting of three nodes with some files each. One of this test cases is given in Figure 3. Class `Test` models a user that communicates with one of the Peers through the user interface and searches for a file document named "f1"; the result is stored in the variable `reply`.

In order to examine the paths generated by this test case, we adapt the class by replacing the constant "f1" with the symbolic variable `reqkeyS` (Figure 4). Recall that DSE performs a concrete and a symbolic run in parallel. The DSE interpreter of Creol therefore treats `reqkeyS` as normal variable for the concrete run, but as symbolic value in the symbolic execution. The assign of the original value in Line 6 is only executed to generate the concrete run, the symbolic execution passes the symbolic value `reqkeyS` to the method `cl.search`. Running the DSE interpreter on this program gives us two decisions in `if` statements within the peers that depend on `reqkeyS`:

```

{"ifthenelse" : not( in(reqkeyS, ["f2"])) }
{"ifthenelse" : in( reqkeyS, ["f1"]) }

```

The conditions represent checks if `reqkeyS` is in the list of files that are stored at a server. The first server (Condition 1) has the file list ["f2"], which does not contain "f1". The concrete run therefore proceeds to the branch of the conditional where `not(in(reqkeyS, ["f2"]))` is true (the `else` branch). The check at the second server is successful (Condition 2). Manual examination of all predefined test cases quickly shows that this pattern repeats for each test case. For proper coverage, we are interested in concrete values of `reqkeyS` not satisfying the already taken decisions. In our example this means that we need a value executing a path that does not end in finding a file. Hence, a new concrete value (e.g. "f0") that is not contained in any of the three servers is assigned to `reqkeyS`, what leads to the following path condition:

```

{"ifthenelse" : not( in(reqkeyS, ["f2"])) }
{"ifthenelse" : not( in(reqkeyS, ["f1"])) }
{"ifthenelse" : not( in(reqkeyS, ["f1", "f2", "f3"])) }

```

This new test case represents the important case that ensures that all servers are contacted and the client performs properly even if no file was found.

6.2 The ASK System

ASK is an industrial software system for connecting and organizing people, developed by the research company Almende and marketed by ASK Community Systems. The ASK system provides mechanisms for matching users requiring information or services with potential suppliers and is used by various organizations for applications like workforce planning and emergency response. The number of people connected varies from several hundred to several thousands.

A Creol reference model for ASK systems has been developed by Almende [1]. The ASK system consists of a number of components to receive and process requests. Each of these components is itself multi-threaded. The threads inside a component act as workers in a thread pool, the executing tasks are put into a component-wide shared task queue. A *balancer* is used to create and destroy worker threads depending on a given maximal number of threads, the currently existing number of threads and on the number of remaining tasks. Figure 5 shows one central part of this balancing task: the tail-recursive method `createThreads`. This method and its opponent in the model, `killThreads`, are responsible for creating and killing threads when appropriate. The balancer is initialized with the symbolic value $maxthreads_S$, the maximum number of threads that are allowed in the thread pool. Inside the balancer, the local variable `maxthreads` is then set to $maxthreads_S + 1$ to account for the balancer thread itself, which also runs inside the thread pool. The balancer has access to the number of threads that are active (`tthreads`), the number of threads that are processing some task (`busythreads`), and the number of tasks that are waiting to be assigned to a worker thread (`tasks`).

The `await` statement in Line 4 suspends the process if it is not necessary to create further worker threads, i.e. if the maximal number of threads is already reached or half of the threads are without a task (they are neither processing a task, nor is there a task open for processing). The `if` statement in Line 7 makes sure there are not more tasks created than allowed by `maxthreads`. Finally, the thread pool is ordered to create the required numbers of threads in Line 11.

We instantiate the model with a fixed number of tasks (10 in our example) and with a variable maximum of threads $maxthreads_S$, with the goal of finding different values for $maxthreads_S$ to optimize the coverage of the code in Figure 5. In the following, we show only the relevant parts of the calculated path conditions, leaving out conditions pertaining to other parts of the model (`killThreads`, the thread creation code inside `threadpool`, etc.).

For a first run we choose $maxthreads_S=0$. Dynamic symbolic execution with this starting value results in the path condition:

```
{"disabled_await" : not( 1 < (maxthreads_S + 1) & true) }
```

After a little simplification it becomes clear that the path was taken because $0 \geq maxthreads_S$. Any other start value will lead to a different run. We select a start value $maxthreads_S=15$ and get

```
{"enabled_await" : (1 < (maxthreads_S + 1) & true) }  
{"ifthenelse" : not(10 > maxthreads_S ) }
```

```

1  op createThreads ==
2    var amountToCreate : Int;
3    var idlethreads : Int := threads - busythreads;
4    await ((threads < maxthreads)
5            $\wedge$  ((idlethreads - tasks) < (threads / 2)));
6    amountToCreate := tasks - idlethreads + (threads / 2);
7    if (amountToCreate > (maxthreads - threads)) then
8      amountToCreate := maxthreads - threads;
9    end;
10   if (amountToCreate > 0) then
11     await threadpool.createThreads(amountToCreate);
12   end;
13   createThreads();

```

Fig. 5. Model of thread pool balancing code in the ASK system. The fields `threads`, `idlethreads` and `tasks` are updated by outside method calls, so the conditions in the `await` statements can become true.

The number 10 reflects the number of tasks we created. The path condition reflects that all inputs with $maxthreads_S \geq 10$ lead to the same path because in each case only the number of threads is created, which is 10 due to the 10 tasks with which the model was initialized. There is no condition for the `if` in Line 10 because the amount to create does not exceed $maxthreads_S$ and therefore is not dependent on it. A third run, created with $maxthreads_S=5$, results in

```

{"disabled_await" : (1 < (maxthreads_S + 1) & true) }
{"ifthenelse" : 10 > maxthreads_S }
{"ifthenelse" : maxthreads_S > 0 }

```

In this test case the amount of tasks to create exceeded the maximal allowed number of tasks and therefore was recomputed in Line 8. The new value depends on $maxthreads_S$, which causes the `if` statement in Line 10 to contribute to the path condition. The new path condition does not further divide the input space, so the maximal possible coverage according to the chosen coverage criterion is reached.

7 Conclusions

The main contribution of this work is the novel extension of dynamic symbolic execution to non-trivial distributed and concurrent object models. This has been achieved by exploiting the properties of the Creol modeling language; in particular local scheduling control of the processes and strict encapsulation of the object state. This paper demonstrates how dynamic symbolic execution, combined with the executable architectural models of Creol, can be used to systematically derive interesting test cases, while avoiding the combinatorial explosion inherent in distributed concurrent systems. Our approach has been formalized in rewriting

logic and implemented in Maude. A peer to peer example and an industrial case study of an agent system serve to illustrate the technique.

The current version of the tool reports the equivalence classes to the user, but does not automatically select and execute new test runs. Immediate future work will be an automation of this process by means of constraint solving techniques. Others have shown that this is feasible in practice, e.g. in [19].

Dynamic symbolic execution, as presented in this paper, should be applicable to other object-oriented languages with concurrency by enforcing serialization of processes in the object as well as strict encapsulation. In a multi-threaded concurrency model as found in Java, dynamic symbolic execution could in principle be achieved by declaring all methods as synchronized and all fields as private. However, such severe restrictions seem undesirable. It would be interesting if lighter restrictions for such languages could be identified that still enable dynamic symbolic execution.

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