Bounded Symbolic Execution for Runtime Error Detection of Erlang Programs *

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Dynamically typed languages, like Erlang, allow developers to quickly write programs without explicitly providing any type information on expressions or function definitions. However, this feature makes those languages less reliable than statically typed languages, where many runtime errors can be detected at compile time. In this paper, we present a preliminary work on a tool that, by using the well-known techniques of metaprogramming and symbolic execution, can be used to perform bounded verification of Erlang programs. In particular, by using Constraint Logic Programming, we develop an interpreter that, given an Erlang program and a symbolic input for that program, returns answer constraints that represent sets of concrete data for which the Erlang program generates a runtime error.

1 Introduction

Erlang [16] is a functional, message passing, concurrent language with dynamic typing. Due to this type discipline, Erlang programmers are quite familiar with typing and pattern matching errors at runtime, which normally appear during the first executions of freshly written programs. Less often, these errors will be undetected for a long time, until the user inputs a particular value that causes the program to crash or, in the case of concurrent programs, determines a particular interleaving that causes an error to occur.

In order to mitigate these problems, many static analysis tools have been proposed. Here let us recall:

- Dialyzer [13], which is a popular tool included in Erlang/OTP for performing type inference based on success typings, and
- SOTER [7], which is a tool that performs verification of Erlang programs by using model checking and abstract interpretation.

However, those tools are not all fully automatic, and they can only be used to cover either the sequential or the concurrent part of an Erlang program, but not both.

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In this paper we present a preliminary work on a technique, based on Constraint Logic Programming (CLP) [10], for detecting runtime errors in Erlang programs. In our approach, sequential Erlang programs are first translated into Prolog facts and then run by using an interpreter written in CLP. Our CLP interpreter is able to run programs on symbolic input data, and it can perform verification of Erlang programs up to a fixed bound on the number of execution steps.

The Erlang language. In this work we consider only sequential programs written in a first-order subset of the Core Erlang language¹. The syntax of this subset can be found in Figure 1.

Figure 1: Language syntax rules

Here, a module is a sequence of function declarations, where each function name has an associated definition of the form "fun $(X_1, ..., X_n)$ -> expr end" (for simplicity, we assume that programs are made out of a single module). The body of a function is an expression expr, which can include variables, literals (atoms, integers, floats, or the empty list), list constructors, tuples, let expressions, case expressions, function applications, calls to built-in functions, and try/catch blocks.

In a case expression "case expr of $clause_1; \dots; clause_m$ end", the expression expr is first reduced to a value v, and then v is matched against the clauses "pat when $expr_1 \rightarrow expr_2$ " of the case expression. The first clause to match this value (i.e., the first clause for which there exists a substitution σ such that $v = pat \sigma$ and $expr_1 \sigma$ reduces to true) is selected, and the computation continues with the evaluation of the clause body (after updating the environment with σ).

Let us remark that primop expressions of the form "primop $atom\ (expr_1, \ldots, expr_n)$ " are primitive operation calls. In general, their evaluation is implementation-dependent, and they may have side effects or raise exceptions. However, in our setting, these are mainly used for raising exceptions in pattern matching errors.

The Erlang program in Figure 2 will successfully compile with no warnings in Erlang/OTP and will correctly compute the sum of the elements in L provided that L is a list of numbers.

¹Core Erlang is the intermediate language used by the standard Erlang compiler, which removes most of the syntactic sugar present in Erlang programs.

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```
-module(sum_list).
-export([sum/1]).
sum(L) ->
  case L of
    [] -> 0;
    [H|T] -> H + sum(T)
  end.
```

Figure 2: A program in Erlang that computes the sum of all numbers in the input list L.

Otherwise, the program generates a runtime error. For instance, if the input to sum is an atom, then the program crashes and outputs a pattern matching error (match_fail), because there are no patterns that match an atom. Similarly, if the input to sum is a list of values, where at least one element is an atom, the execution halts with a type error (badarith), when applying the function '+' to a non-numerical argument.

The tool Dialyzer does not generate any warnings when analyzing this program. That is coherent with the Dialyze approach, which only complains about type errors that would guarantee the program to crash for all input values. However, it might be the case that we want to perform a more detailed

analysis on this program. In the following, we will see how our tool lists all the potential runtime errors, together with the input types that can cause them.

2 Symbolic Interpreter for Runtime Error Detection of Erlang Programs

The main component of the verifier is a CLP interpreter that defines the operational semantics of our language. This executable specification of the semantics enables the execution of Erlang programs represented as Prolog facts.

Therefore, we have defined a translation from Erlang programs to Prolog facts. More precisely, our translator generates one fact fundef for each function definition occurring in the Core Erlang code obtained from the original Erlang module. For instance, the translation of the function sum/1 defined in the module sum_list corresponds to the fact which can be seen in Figure 3.

Figure 3: Prolog fact generated by the Erlang-to-Prolog translation for the sum function definition.

This translation is quite straightforward, since the standard compilation from Erlang to Core Erlang greatly simplifies the code. Note that, since we generate Prolog facts, we have used the cons predicate for Erlang list constructors to distinguish them from Prolog list constructors. Note also that an additional *catch-all* clause has been added for the case in which the argument does not match any of the clauses from the case expression (i.e., for pattern matching errors). This transformation and similar ones are automatically made by the standard compilation from Erlang to Core Erlang.

The CLP interpreter provides a flexible means to perform the *bounded verification* of Erlang programs. By using a symbolic representation of the input data, the interpreter allows the exhaustive exploration of the program computations without explicitly enumerating all the concrete input values. In particular, the interpreter can run on input terms containing logic variables, and it uses constraint solvers to manipulate expressions with variables ranging over integer or real numbers. By fixing a bound to limit the number of computation steps performed by the interpreter, we force the exploration process to be finite, and hence either we detect a runtime error or we prove that the program is error-free up to the given bound.

Let us consider an Erlang program Prog, which is translated into Prolog facts as shown in Figure 3. In order to perform the bounded verification of the Erlang program Prog, the interpreter provides the predicate run(FName/Arity,Bound,In,Out), whose execution evaluates the application of the function FName of arity Arity to the input arguments In in at most Bound steps. The arguments In are represented as a Prolog list (written using square brackets) of length Arity. Out is the result of the function application. If the evaluation of the function application generates an error, then Out is bound to a term of the form error(Err), where Err is an error name (e.g., match_fail, indicating a match failure, or badarith, indicating an attempt to evaluate an arithmetic function on a non-arithmetic input), meaning that the specific error Err had occurred. Hence, the bounded verification of a given Erlang program can be performed by executing a query of the form:

```
?- run(FName/Arity, Bound, In, error(Err)).
```

where FName is a constant, Arity and Bound are non-negative integers, and In and Err are, possibly non-ground, terms.

Any answer to the query is a successful detection of the error Err generated by evaluating the application of the function FName to the input In. If no answer is found, then it means that no error is generated by exploring the computation of FName up to the value of Bound, and we say that the program Prog is correct up to the given bound.

Now let us see the bounded verifier in action by considering the sum_list program of the previous section and the following query:

```
?- run(sum/1,20,In,error(Err)).
```

Among the answers to the query, we get the following constraints relative to the input In and the error Err:

```
In=[cons(lit(Type,_V),lit(list,nil))],
Err=badarith,
dif(Type,int), dif(Type,float)
```

meaning that if sum/1 takes as input a list (represented as a Prolog term of the form cons(Head, Tail)) whose head is not a numeric literal (denoted by the constraints dif(Type,int) and dif(Type,float)), then a badarith error occurs, that is, a non-numerical argument is given as input to an arithmetic operator. Another answer we get is:

```
In=[L],
Err=match_fail,
dif(L,cons(_Head,_Tail)), dif(L,lit(list,nil))
```

meaning that if sum/1 takes as input an argument L which is neither a cons nor a nil term, then a match_fail error occurs. Note that, due to the recursive definition of sum, the bound is essential to detect this error.

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Now let us introduce the predicate int_list(L,M) that generates lists L of integers of length up to M. For instance, the query

the bounded verifier terminates after 0.347 seconds² with answer false, meaning that if the input to sum is any list of 100 integers, then the program is correct up to the bound 100. Note that no concrete integer element of the list is needed for the verification of this property.

Now we sketch the implementation of the operational semantics of Erlang expressions. The semantics is given in terms of a transition relation of the form tr(Bound, ICfg, FCfg), which defines how to get the final configuration FCfg from the initial configuration ICfg in Bound computation steps. Configurations are pairs of the form cf(Env, Exp), where Env is the environment mapping program variables to values and Exp is a term representing an Erlang expression.

The environment is extended with a boolean flag that keeps track of the occurrence of any runtime error during program execution. The value of the error flag Flag in the environment Env can be retrieved by using the predicate lookup_error_flag(Env,Flag). The value of the flag in a given environment EnvIn can be updated by using the predicate update_error_flag(EnvIn, Flag,EnvOut), thereby deriving the environment EnvOut whose error flag is set to Flag. If the evaluation of IExp generates the error Err, then FExp is a term of the form error(Err) and the error flag is set to true.

In Figure 4 we present the clause for tr/3 that implements the semantics of function applications represented using terms of the form apply(FName/Arity,IExps), where FName is the name of a function of arity Arity applied to the expressions IExps. The transition performed by tr/3 only applies if:

- 1. no error has occurred so far, that is, lookup_error_flag(IEnv,false), and
- 2. the bound has not been exceeded, that is, Bound > 0.

Then, the following operations are performed:

- 3. the value of the bound Bound is decreased,
- 4. the parameters FPars and the body FBody of the function FName of arity Arity are retrieved (the predicate lookup_fun/3 is responsible for extracting FPars and FBody from the fundef fact representing FName/Arity),
- 5. the list of the actual parameters IExps is evaluated in IEnv, thereby deriving the list of expressions EExps and the new environment EEnv (it may differ from IEnv in the error flag and new variables occurring in the expressions IExps may have been added),
- 6. the formal parameters FPars are bound to the expressions IExps, thereby deriving the new environment AEnv,
- 7. the error flag in AEnv is updated to the value F1 in EEnv, thereby deriving the environment BEnv,

 $^{^2}$ The query has been executed using SWI-Prolog v7.6.4 (http://www.swi-prolog.org/) on an Intel Core i5-2467M 1.60GHz processor with 4GB of memory under GNU/Linux OS

```
tr(Bound,cf(IEnv,IExp),cf(FEnv,FExp)) :-
      IExp = apply(FName/Arity,IExps),
      lookup_error_flag(IEnv,false),
                                                 % 1
      Bound > 0,
                                                 % 2
      Bound1 is Bound - 1,
                                                 % 3
      lookup_fun(FName/Arity,FPars,FBody),
                                                 % 4
      tr_list(Bound1, IEnv, IExps, EEnv, EExps),
                                                 % 5
      bind(FPars,EExps,AEnv),
                                                 % 6
      lookup_error_flag(EEnv,F1),
      update_error_flag(AEnv,F1,BEnv),
                                                 % 7
      tr(Bound1,cf(BEnv,FBody),cf(CEnv,FExp)),
                                                 % 8
      lookup_error_flag(CEnv,F2),
      update_error_flag(EEnv,F2,FEnv).
                                                 % 9
```

Figure 4: Definition of the operational semantics for a function application apply/2.

- 8. the body FBody is evaluated in BEnv to get the final expression FExp, and
- 9. the final environment FEnv is obtained from EEnv by setting the error flag to the value F2 obtained from the callee function.

Each rule of the operational semantics for Erlang programs is translated into a clause for the predicate tr/3. These clauses are omitted.

Now we can present the definition of run/4, which depends on tr/3:

```
run(FName/Arity,Bound,In,Out) :-
lookup_fun_pars(FName/Arity,FPars),
bind(FPars,In,IEnv),
tr(Bound,cf(IEnv,apply(FName/Arity,FPars)),cf(FEnv,Out)).
```

The predicate run retrieves the formal parameters FPars of FName/Arity and creates an environment IEnv where those parameters are bound to the input values In. Then, it evaluates the application of FName to its parameters, thereby producing the final expression Out.

3 Conclusions and future work

We have presented a work in progress for the development of a CLP interpreter for detecting runtime errors of Erlang programs. An Erlang program is first translated into a set of Prolog facts, then the CLP interpreter is run using this translation together with symbolic input data. At present, our interpreter is able to deal with first-order sequential Erlang programs, but we think that the extension to higher-order functions can be achieved by following a similar approach. In the future, we also plan to consider concurrency with an appropriate technique for handling the state explosion problem. For instance, we may employ a partial order reduction technique [1] to obtain the minimal set of concurrent behaviours for a given program, and then generate the associated executions using our interpreter.

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Let us briefly compare our work with the static analysis tools available for Erlang. Unlike Dyalizer [13], our tool computes answer constraints that describe type-related input patterns which lead to runtime errors. However, as already mentioned, due to the bounded symbolic execution, our interpreter may terminate with no answer, even if runtime errors are possible for concrete runs which exceed the given bound. One of the weaknesses of Dialyzer is that it is hard to know where typing errors come from. An extension of Dialyzer that provides an explanation for the cause of typing errors has been proposed to overcome this problem [15]. We believe that we are able to provide a similar information if we include debugging information in the clauses generated by our Erlang-to-CLP translation.

Unlike SOTER [7], which is based on abstract interpretation, our CLP interpreter provides full support to arithmetic operations through the use of constraint solvers. Moreover, the symbolic interpreter does not require any user intervention (except for the bound), while in SOTER the user is responsible for providing a suitable abstraction.

Besides being useful on its own for bounded verification, the CLP interpreter for Erlang may be the basis for more sophisticated analysis techniques. In particular, by following an approach developed in the case of imperative languages, we intend to apply CLP transformation techniques to specialize the interpreter with respect to a given Erlang program and its symbolic input [6]. The specialized CLP clauses may enable more efficient bounded verification, and they can also be used as input to other tools for analysis and verification (such as constraint-based analyzers [3, 11] and SMT solvers [8, 14]), which have already been shown to be effective in other contexts [2, 4, 5]. Moreover, the specialized clauses can be used to apply backward analysis techniques for CLP programs based on abstract interpretation (see, for instance, [9, 12]). Backward analysis aims at deriving from a property that is expected to hold at the end of the execution of a program, conditions on the query which guarantee that the desired property indeed holds. In our context, backward analysis can be applied to deduce those conditions that may cause a runtime error, and then use them to improve the forward symbolic execution of the program.

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