Automated Test Generation and Verified Software*

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Abstract. Testing remains the principal means of verification in commercial practice and in many certification regimes. Formal methods of verification will coexist with testing and should be developed in ways that improve, supplement, and exploit the value of testing. I describe automated test generation, which uses technology from formal methods to mechanize the construction of test cases, and discuss some of the research challenges in this area.

1 Introduction

By *testing* I mean observation of a program in execution under controlled conditions. Observations are compared against an explicit or informal oracle to detect bugs or confirm correctness. Much of the testing process (i.e., the execution and monitoring of tests) is automated in modern development environments, but construction of test cases (i.e., the specific experiments to be performed) remains a largely manual process.

Testing is the method by which most software is verified today. This is true for safety critical software as well as the commodity variety: the highest level of flight critical software (DO-178B Level A) is required to be tested to a structural code coverage criterion known as MC/DC (Modified Condition/Decision Coverage) [1]. And although formal methods of analysis and verification are becoming sanctioned, even desired, by some certification regimes, testing continues to be required also—because it can expose different kinds of problems (e.g., compiler bugs), can examine the program in its system context, and increases the diversity of evidence available.

The weakness of testing is well-known to the formal methods and verification communities—it can only show the presence of bugs—but those communities are now beginning to recognize its strength: it *can* show the presence of bugs—often, very effectively. It is a great advantage in verification if the software to be verified is actually correct, so inexpensive methods for revealing incorrectness early in

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the development and verification process are necessary for verified software to be economically viable.

Thus, testing is not a rival to formal methods of verification but a valuable and complementary adjunct. It is worthwhile to study how each can support the other, both in the technology that they employ, and in their contribution to the overall goal of cost-effective verification.

In this regard, the most significant recent development in testing has been the application of technologies from verification (notably, model-checking, SAT and SMT solving, and constraint satisfaction) to automate the generation of test cases. Automated test generation poses urgent challenges and opportunities: there are many technical challenges in achieving effective automation, there is a wealth of opportunity in the different ways that automated testing can be used, and there are serious implications for traditional certification regimes—and opportunities for innovative ones; there are also opportunities for theoretical research in the relationship between testing and verification, and for empirical inquiry into their pragmatic combination.

In this paper, I briefly survey the topics mentioned above, and suggest research directions for the development and use of automated test generation in verification.

2 Technology for Automated Test Generation

Much of the process of test execution and monitoring is automated in modern software development practice. But the generation of test cases has remained a labor-intensive manual task. Methods are now becoming available that can automate this process.

A simple test-generation goal is to find an input that will drive execution of a (deterministic, loop-free) program along a particular path in its control flow graph. By performing symbolic execution along the desired path and conjoining the predicates that guard its branch points, we can calculate the condition that the desired test input must satisfy. Then, by constraint satisfaction, we can find a specific input that provides the desired test case. This method generalizes to find tests for structural coverage criteria such as statement or branch coverage, and for programs with loops and those that are reactive systems (i.e., that take an input at each step). A major impetus for practical application of this approach was the realization that (for finite state systems) it can be performed by an off-the-shelf model checker: we simply check the property "always not P," where P is a formula that specifies the desired structural criterion, and the counterexample produced by the model checker is then the test case desired [2]. Different kinds of structural or specification-based tests can be generated by choosing suitable P.

Using a model checker to generate tests in this way can be very straightforward in model-based development, where we have an executable specification for the program that is in, or is easily translated to, the language of a model checker: the tests are generated from the executable specification, which then provides the oracle when these are applied to the generated program. There are many pragmatic issues in the selection of explicit-state, symbolic, or bounded

model checkers for this task [3] and it is, of course, possible to construct specialized test generators that use the technology of model checking but customize it appropriately for this application.

The test generation task becomes more challenging when tests are to be generated directly from a low-level program description, such as C code, when the path required is very long (e.g., when it is necessary to exhaust a loop counter), when the program is not finite state, and when nondeterminism is present.

When a higher-level specification is unavailable and tests must be generated directly from C code or similar low-level description, it is natural to adopt techniques from software model checking. These seldom translate the program directly into the language of a model checker but usually first abstract it in some way. Predicate abstraction [4] is the most common approach, and discovery of suitable predicates is automated very effectively in the lazy-abstraction approach [5]. Abstractions for test generation are not necessarily the same as those used for verification. For the latter, the abstraction needs to be conservative (i.e., it should have more behaviors than the concrete program), whereas in the former case we generally desire that any test generated from the abstraction should be feasible in the concrete program (i.e., the abstraction may have fewer behaviors than the concrete program) [6]. This impacts the method for constructing the abstraction, and the choice of theorem proving or constraint satisfaction methods employed [7].

When very long test sequences are needed to reach a desired test target, it is sometimes possible to generate them using specialized model checking methods (e.g., those based on an ATPG engine [8]), or by generating the test incrementally, so that each subproblem is within reach of the model checker [3]. Some of the most effective current approaches for generating long test sequences use combinations of methods. For example, random test generation rapidly produces many long paths through the program; to reach an uncovered test target, we find a location "nearby" (e.g., measured by Hamming distance on the state variables) that has been reached by random testing and then use model checking or constraint satisfaction to extend the path from that nearby location to the one desired [9]. DART [10] uses a different approach to explore all feasible execution paths: the program under test is first executed on some random input and monitored to gather constraints on inputs at conditional branches during that run; then a constraint solver is used to generate variants on the inputs to steer the next execution of the program towards different execution paths.

Traditional model checking technology must be extended or adapted when the program is not finite state. In some cases, an infinite state bounded model checker can be used (i.e., a bounded model checker that uses a decision procedure for satisfiability modulo theories (SMT) [11] rather than a Boolean SAT solver) [12]. An SMT solver can, for example, generate real-valued inputs that can be interpreted as delays to be used in testing a real-time program.

In cases where inputs to the program are not simple numerical quantities but data types such as trees or lists, a plausible approach is to fix the base type (e.g., elements of the tree are chosen from an alphabet of size 2), bound

the size of the data type (e.g., trees with no more than 5 nodes), and then generate all or many instances of the data type within those constraints. This is easily automated when an axiomatic specification for the data type is available (e.g., rewrite rules specifying a tree), but straightforward approaches produce highly redundant tests (i.e., they generate many inputs that are structurally "isomorphic" to each other). Gaudel and her colleagues have developed methods for generating tests in this way while reducing redundancy using "regularity" and "uniformity" hypotheses [13, 14].

A variant is where we lack a generator for the data type but have a recognizer for it: for example, we may have a predicate that recognizes red-black trees represented as linked lists. Here, we could randomly or exhaustively generate all inputs up to some specified size and test them against the recognizing predicate, but this is very inefficient (e.g., very few randomly generated list structures represent a valid red-black tree) and also generates many "isomorphs." Hence, it is best to view the search as a constraint satisfaction problem and to use technology from that domain [15].

The test generation problem changes significantly when the program under test is nondeterministic, or when part of the testing environment is not under the control of the tester (e.g., testing an embedded system in its operational environment). In these cases, we cannot generate test sequences independently of their actual execution: it is necessary to observe the behavior of the system in response to the test generated so far and to generate the next input in a way that advances the purpose of the test (and to recognize when this cannot be achieved and the current test should be abandoned). This kind of testing can be seen as a game between the tester and the system under test: rather than passive test cases (i.e., data) we need an active tester (i.e., a program), and test generation becomes a problem of controller synthesis. Methods for solving this problem can use technology similar to model checking but can seldom use an off-the-shelf model checker [16]; SpecExplorer is a tool of this kind [17].

The problem becomes yet more difficult when the test environment includes mechanical or biological systems: for example, testing the shift controller of an automatic gearbox in its full system context with a (real or simulated) gearbox attached, or testing a pacemaker against a simulated heart. Here, the test generation problem is escalated to one of controller synthesis in a hybrid system (i.e., one whose description includes differential equations). This is a challenging problem, but a plausible approach is to replace the hybrid elements of the modeled environment by conservative discrete approximations, and then use methods for test generation in nondeterministic systems [18]. As in the case of predicate abstraction, the notion of "conservative" that is suitable for test generation may differ from that used in verification.

3 Selection of Test Targets

The previous section has sketched how test cases can be generated automatically; the next problem is to determine how to make good use of this capability.

One approach uses test generation to help developers explore their emerging designs [19]: a designer might say "show me a run that puts control at this point with $x \leq 0$." This approach is very well-suited to model-based design environments (i.e., those where the design is executable), but is less so for traditional programming. An approach that has proven useful in traditional programming is random test generation at the unit level. In some programming environments, each unit is automatically subjected to random testing against desired properties if these have been specified, or generic ones (e.g., no exceptions) as it is checked in (Haskell QuickCheck [20] is the progenitor of this approach). A similar approach can be used in theorem proving environments: before attempting to prove a putative theorem, first try to refute it by random test generation [21] (in PVS, this can also be tried during an interactive proof if the current proof goal looks intractable [22]). These simple approaches are highly effective in practice. More challenging tests can be achieved by exhaustive generation of inputs up to some bounded size [23]. In Extreme Programming, tests take on much of the rôle played by specifications in more traditional development methods [24], and automated, incremental test generation can support this approach [25].

More traditional uses of testing are for systematic debugging, and for validation and verification. In tests developed by humans, the first of these is generally driven by some explicit or implicit hypotheses about likely kinds of bugs, while the others are driven by systematic "coverage" of requirements and code.

One simple fault hypothesis is that errors are often made at the boundaries of conditions (e.g., the substitution of $<$ for $\leq$) and some automated test generators target these cases [26]. Another hypothesis is that compound decisions (e.g., $A \wedge B \vee C$) may be constructed incorrectly so tests should target the "meaningful impact" [27] of each condition within the decision (i.e., each must be shown able to independently affect the outcome).[1] It turns out that these ideas are related: boundary testing for $x \leq y$ is equivalent to rewriting the decision as $x < y \vee x = y$ and then testing for meaningful impact of the two conditions. The classes of faults detected by popular test criteria for compound decisions have been analyzed by Kuhn [28] and extended by others [29, 30].

Requirements- or specification-based testing is most easily automated when the requirements or specification are provided in executable form—as is commonly done in model based development, thereby giving rise to model-based testing [31]. Here, we can use the methods sketched in Section 2 to generate tests that systematically explore the specified behavior. The usual idea is that a good set of tests should thoroughly explore the control structure of the specification; typical criteria for such structural coverage are to reach every control state, to take every transition between control states, and more elaborate variants that explore the conditions within the decisions that control selection of transitions (as in the meaningful impact criteria mentioned earlier). Structural coverage criteria can be augmented by "test purposes" [32] that describe the kind of tests we want to generate (e.g., those in which the `gear` input to a gearbox shift se-

[1] This use of *decision* and *condition* is the one employed in MC/DC, which is a testing criterion of this kind.

lector changes at each step, but only to an adjacent value), by predicates that describe relationships that should be explored (e.g., a queue is empty, full, or in between) [33], or by specifications that describe the external environment [34]. Test purposes and predicates are related to predicate abstraction and can be used to reduce the statespace of the model, and thereby ease the model checking task underlying the test generation. Generating a separate test for each coverage target produces inefficient test sets that contain many short tests and much redundancy, so recent methods attempt to construct more efficient "tours" that visit many targets in each test [35, 3, 33].

Requirements-based testing is more difficult when requirements are specified by properties rather than models. One approach is to translate the properties into automata (i.e., synchronous observers), then target structural coverage in the automata. Alternatively, a direct approach is described in [36].

4 Testing for Verification

Certification regimes for which testing is an important component generally require evidence that the testing has been thorough. DO-178B Level A (which applies to the highest level of flight-critical software in civil aircraft) is typical: it requires MC/DC code coverage. The expectation is that tests are generated by consideration of requirements and their execution is monitored to measure coverage of the code. As the industry moves toward model-based development, it can be argued that the requirements are represented by the models, and hence that automated test generation from the model is a form of requirements-based testing. One way to do this is by targeting MC/DC coverage in the model. Heimdahl, George, and Weber did this for a model of a flight guidance system developed by Rockwell, and then executed the tests on implementations that had been seeded with errors [37]. They found that the autogenerated tests detected relatively few bugs, and generally performed worse than random testing. Part of the explanation for this distressing observation is that the model checking technology underpinning the test generation is "too clever": it generally finds the *shortest* test to discharge any given goal, and these short tests often exploit some special case and never reach the interesting parts of the state space. There is hope that methods that generate tours through many test goals will do better than those that target the goals individually, or that suitable test purposes may guide the test generator into more productive areas of the state space, but these ideas need to be validated in practice.

Another way in which testing has been employed for verification is in "conformance testing," which is generally applied to distributed systems and protocols (where the tester must be an active program). Given a formal specification and an implementation that purports to satisfy it, conformance testing generates a series of tests such that any departure from the specification will eventually be revealed (subject to various technical caveats) [38]. Only a relatively small number of tests can be performed in practice, so the eventuality guarantee is of mainly theoretical interest, and the more pragmatic concern is to try and arrange things so that tests generated early in the series are effective at finding bugs.

Until recently, there has been relatively little work that uses automated testing and static analysis or formal verification in combination. One idea is to reduce "false alarms" in static analysis by emitting only those errors for which a test case to manifest it can be constructed. Dually, some static analysis methods (such as slicing) can be used to ease the test generation task [39]. Rusu uses test generation to decompose the classical formal verification problem into smaller components [40], while the Synergy method [41] alternates DART-like testing and formal verification to achieve highly automated property checking.

5 Research Challenges

Testing is the dominant means of verification used today. Any research agenda in software verification must include testing as a topic, and its roadmap must suggest how the proposed research will improve testing, and how it can use it, as well as how it may replace it in selected areas.

Automated test generation is an attractive topic in this area: it can reduce the cost of testing and may improve its quality. And it is an "invisible" application of formal methods and thus provides a good opportunity to introduce this technology to new communities. Among the most eager adopters of this capability are those in regulated industries where onerous testing requirements constitute a significant part of overall development costs.[2] As mentioned above, there is some evidence that simply using the test coverage requirements as a target for automated test generation may be a flawed strategy: coverage metrics are intended to measure the thoroughness of human-generated tests, and do not necessarily lead to good test sets when used in an inverted role as a specification for the tests required.

Thus, an urgent research topic is development of techniques for specifying good test sets. There are two subtopics here: the role of the human tester will change from construction of tests to *specification* of tests (the tests will be generated automatically from the specification), so we need ideas and techniques for specifying tests (e.g., an extended notion of test purpose); second, we need empirical data on what kinds of test specification produce good tests (i.e., those that are effective in revealing errors). It is natural to assume that specifications for tests should directly correspond to a method for generating the tests, but modern automated test generation operates by constraint satisfaction (either explicitly, or implicitly via model checking) and this opens up the possibility that tests can be specified indirectly in terms of recognizers. Concretely, we can specify tests indirectly by means of synchronous observers that "recognize a good test when they see one" and raise a flag to indicate it. Then we use a model checker to find circumstances that raise the flag. This use of recognizers creates many attractive possibilities for test specification [43].

Most current methods and tools for automated test generation are limited to unit tests. A second general research area is development of methods and

[2] Alternatively, there has been some industrial use of formal verification as a lower-cost replacement for MC/DC testing [42].

technology for other (arguably more important) testing tasks, such as integration and system tests. At these levels, tests become interactive programs, and the formal context becomes that of controller synthesis for nondeterministic, timed, and hybrid systems. Abstraction is likely to be necessary, both for the system under test and for its environment, and there are interesting questions regarding the appropriate kinds of abstractions to use, and the theorem proving and model checking methods that are most suitable for constructing and using them. Integration testing need not be restricted to the later stages of development: much of requirements engineering is concerned with anticipation of interactions among components, and the earlier and more completely these can be understood the better. Model checking and automated integration testing during development of (model-based) requirements could reduce the well-known "explosion" of problems that traditionally arise at system integration time.

A third suggested general research area is the integration of testing with formal methods of analysis and verification. Again, there are two subtopics: one is technical integration—for example, how can testing help in formal specification and proof (cf. QuickCheck-like methods for rapid refutation)—while the other focuses on how the overall verification process can be decomposed into elements that are effectively tackled by different means. For example, formal verification not only provides guarantees, it exposes assumptions—and these assumptions can be useful targets for testing. Penetration testing for security at the highest "evaluation assurance levels" uses exactly this approach. Here, the relationship between formal verification and testing is clear, but in other contexts it can be less so. There are proposals, for example, to replace some unit test requirements in avionics by static analysis; yet testing can address some issues (such as compiler bugs, which are a genuine problem) that static analysis does not (unless applied to machine code), so the overall web of argument in support of verification may become interestingly complex.

A companion paper in these proceedings outlines some of the issues in technical integration of verification components [44], while the larger issues of "multi-legged assurance," in which the assurance case for a system is composed from different kinds of evidence, is only just beginning to receive attention [45, 46, 47] and is a worthy topic for future study.

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A Discussion on John Rushby's Presentation

Greg Nelson

John, I enjoyed your talk, especially the proposal part of it, and I have two amendments to suggest to your proposal. First of all, you said that you did not see any reason not to have the top-level logic to be quite comprehensive, a superset of all the logics used by the tools, and you then rapidly read a paragraph of features including, I remember, dependent typing but I heard many others. And that caused alarm bells to go off in my mind, because I found that it is easier to integrate a large number of features into a logic than into a programming language, and it is not trivial. You can't always get all the features that you want to be added. For example, it is very difficult to combine tools that assume a typed logic with an untyped logic. And if you throw in dependent subtyping, because you know that it is useful some time, you are actually excluding many tools from the tool bus.

John Rushby

Yes, that is true. So, for example, SAL has predicate subtypes like PVS, which does not have full dependent subtyping. [*Missing sentence*]. So I think there is some discussion possible here, mainly among those who can go into the highly technical details. I'd observe, some of what is going on on the bus is typing judgements. You know whether a thing is a finite state machine and that itself is a predicate subtype, whether it is dependent or not is up to the discussion.

Greg Nelson

I just was conveying my concern for what it is worth to you that I would try to make that top thing as simple as possible rather than as comprehensive as possible.

John Rushby

Okay, we got 15 years of competition to figure this out.

Greg Nelson

My second proposal is that the query language by which the front-end or top looks for tools on the tool bus to answer queries, I think, will become quite more complicated than what you are describing. Let me give you an example to explain that issue that, I think, is probably most important: Your example was: I want to know if this is a well-typed term and one of the tools that can test well-typedness comes back and returns the answer. It may very well be that no single tool can do that, but that one tool can put that query into a certain normal form and another query can find the well-typedness of that normal from, so that you really need to find sequential compositions or functional compositions of tools.

John Rushby

That is exactly right. I glossed over [*some aspects*]. But I did have a slide that talked about how you could discover that you could verify an infinite state system by making it finite state. The same chaining on queries could perhaps build those sequential compositions for tasks like type checking. So I think, your question shows that there are a lot of opportunities for spending money in this direction, and I look forward to those of us trying it.

Tony Hoare

I hope, you will accept *payment* in kind.

John Rushby

Contributions? Certainly. Of tools and code, yes.