Efficient State Merging in Symbolic Execution
(Extended Abstract)

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Recent tools [CDE08, GLM08, CKC11] have applied symbolic execution to automated test case generation and bug finding with impressive results—they demonstrate that symbolic execution brings unique practical advantages. First, such tools perform dynamic analysis and actually execute a target program, including any external calls; this broadens their applicability to many real-world programs. Second, like static analysis, these tools can simultaneously reason about multiple program behaviors. Third, symbolic execution is fully precise, so it generally does not have false positives.

While recent advances in SMT solving have made symbolic execution tools significantly faster, they still struggle to achieve scalability due to path explosion: the number of possible paths in a program is generally exponential in its size. States in symbolic execution encode the history of branch decisions (the path condition) and precisely characterize the value of each variable in terms of input values (the symbolic store), so path explosion becomes synonymous with state explosion. Alas, the benefit of not having false positives in bug finding comes at the cost of having to analyze an exponential number of states.

State merging. One way to reduce the number of states is to merge states that correspond to different paths. Consider, for example, the program if (x<0) {x=0;} else {x=5;} with input X assigned to x. We denote with (pc, s) a state that is reachable for inputs obeying path condition pc and in which the symbolic store \( s = [v_0 = e_0, \ldots, v_n = e_n] \) maps variable \( v_i \) to expression \( e_i \), respectively. In this case, the two states \( (X < 0, [x = 0]) \) and \( (X \geq 0, [x = 5]) \), which correspond to the two feasible paths, can be merged into one state \( (\text{true}, [x = \text{ite}(X < 0, 0, 5)]) \). Here, \( \text{ite}(c, p, q) \) denotes the if-then-else operator that evaluates to \( p \) if \( c \) is true, and to \( q \) otherwise.

State merging effectively decreases the number of paths that have to be explored [God07, HSS09], but also increases the size of the symbolic expressions describing variables. Merging introduces disjunctions, which are notoriously difficult for SMT solvers. Merging also converts differing concrete values into symbolic expressions, as in the example above: the value of \( x \) was concrete in the two separate states, but symbolic \( \text{ite}(X < 0, 0, 5) \) in the merged state. If \( x \) were to appear in branch conditions or array indices later in the execution, the choice of merging the states may lead to more solver invocations than without merging. This combination of larger symbolic expressions and extra solver invocations.

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can drown out the benefit of having fewer states to analyze, leading to an actual decrease in the overall performance of symbolic execution [HSS09].

Furthermore, state merging conflicts with important optimizations in symbolic execution: search-based symbolic execution engines, like the ones used in test case generators and bug finding tools, employ search strategies to prioritize searching of “interesting” paths over “less interesting” ones, e.g., with respect to maximizing line coverage given a fixed time budget. To maximize the opportunities for state merging, however, the engine would have to traverse the control flow graph in topological order, which typically contradicts the strategy’s path prioritization policy.

Our solution. In this work (published as [KKBC12]), we describe a solution to these two challenges that yields a net benefit in practice. We combine the state space reduction benefits of merged exploration with the constraint solving benefits of individual exploration, while mitigating the ensuing drawbacks. Our main contributions are the introduction of query count estimation and dynamic state merging. Query count estimation is a way to statically approximate the number of times each variable will appear in future solver queries after a potential merge point. We then selectively merge two states only when we expect differing variables to appear infrequently in later solver queries. Since this selective merging merely groups paths instead of pruning them, inaccuracies in the estimation do not hurt soundness or completeness. Dynamic state merging is a merging algorithm specifically designed to interact favorably with search strategies. The algorithm explores paths independently of each other and uses a similarity metric to identify on-the-fly opportunities for merging, while preserving the search strategy’s privilege of dictating exploration priorities.

Experiments on all 96 GNU COREUTILS show that employing our approach in a symbolic execution engine achieves speedups over the state of the art that are exponential in the size of symbolic input, and can cover up to 11 orders of magnitude more paths. Our code and experimental data are publicly available at http://cloud9.epfl.ch.

References


