

# A Context-Sensitive Structural Heuristic for Guided Model Checking

Neha Rungta and Eric Mercer Brigham Young University, Provo, UT

### 1. Introduction

 The use of embedded systems has become ubiquitous in recent times · Growing complexity makes ad-hoc testing techniques insufficient · Model checking exhaustively searches program states as shown in Fig. 1 · Computation resources often run out before finishing the search · Guided search ranks states in order of interest with a heuristic Guided search explores interesting states first as in Fig. 2 . The goal is to find an error before running out of computation resources



Fig 1: Breadth First Search

 Breadth-first search orders the search frontier in a FIFO as shown in Fig. 1 Guided search orders the frontier in a priority queue as shown in Fig. 2 The heuristic values on the states is the estimated distance to an error Explore states estimated closer to an error first

#### 2. FSM Distance Heuristic

· FSM builds an Interprocedural Control Flow Graph (ICFG) statically · A depth-first traversal of a program (Fig. 3) builds the ICFG shown in Fig. 4 · FSM maps the current state to a vertex in the ICFG as shown in Fig. 5 · It computes shortest path from a current state to an error state on the ICFG . The length of the shortest path is returned as the heuristic estimate





3. Augmented Interprocedural Control Flow

An AICFG includes a bounded (k) calling context of function calls

Graph (AICFG)

The caller of foo can be resolved by the additional context information There is no underestimation in the path computed in Fig. 8 · Distance estimate computed on an AICFG is more accurate than the FSM For large k's the size of the AICFG can become prohibitive

# 4. Accuracy depends on k-bound

 The context information in the AICFG is limited by the bound • The AICFG for the program in Fig. 9, with a k=1 is shown in Fig. 11 . The reduced call graphs in Fig. 10 shows the lost context in function g



# 5. Extended FSM Distance Heuristic

• The extended FSM (EFSM) distance heuristic recovers calling context as shown in Fig. 12 by using the return addresses for function calls in the runtime stack

. The EFSM takes the runtime stack and extracts AICFG vertices as shown in Fig. 13 that represent the call trace to the current point of execution



• The EFSM heuristic does a forward analysis from the current state (q0) on the AICFG in Fig. 14 to see if an error is reachable within the scope of the function g

• If an error is not reachable, or to find a shorter distance to the error, the EFSM heuristic unrolls the call trace by a single call to q1 in Fig. 12

• The EFSM heuristic again does a forward analysis on the AICFG in Fig. 14 at q2 on all unique paths in the function f that might lead to an error

Repeat the pattern of unrolling the call trace and then doing a forward analysis on the AICFG until the call trace is exhausted

• The EFSM combines the dynamic information on the runtime stack in Fig. 13 with the static information on the AICFG in Fig. 14 to get a better heuristic estimate

## 6. Results

#### States Generated before Error Discovery



· A visible decrease in number of states · FSM guided search degenerates into a random search in some models

## 7. Conclusions

EFSM is admissible and consistent

· EFSM results in a better estimate of the distance to the error · Explores fewer states before error discovery for a set of benchmark examples · EFSM does better in models with densely connected transition graphs · Extra memory used for building and storing the AICFG is negligible Structural heuristics make error discovery more tractable · EFSM can be improved further by using data flow information

#### Time in seconds before Error Discovery



 Overhead in following call trace · EFSM guided search is slower even though it expands fewer states Optimizes space over time · Future work looks at the analysis overhead of the algorithm

#### **Contact Information**

Verification and Validation Laboratory Computer Science Department Brigham Young University Provo, UT 84602

Neha Rungta, neha@byu.edu Eric Mercer, egm@cs.byu.edu http://vv.cs.bvu.edu/