Testing the Temporal Behavior of Real-Time Software Modules using Extended Evolutionary Algorithms

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1 Motivation

Costs for development and test of embedded systems

- \rightarrow 50 % implementation, 30 % unit testing, 20 % system testing
- \Rightarrow 50 % of development costs spent for testing

Objectives of testing real-time systems

- \rightarrow finding errors in functional behavior
- \rightarrow finding errors in temporal behavior
- \rightarrow building up confidence in the correct functioning of the test object by executing the system under test with selected inputs

2 Testing temporal behavior

Testing temporal system behavior

- \rightarrow find violations of specified timing constraints (find the inputs with longest and shortest execution time)
- \rightarrow check whether they produce a temporal error (outputs are produced prematurely or their computation takes too long)
- \Rightarrow testing temporal behavior is a complex task

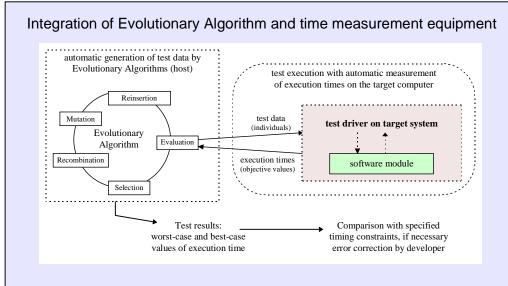
Problem:

- \rightarrow lack of appropriate test procedures
- \rightarrow tester uses conventional test methods



Use of Evolutionary Optimization to determine longest and shortest execution times automatically.

3 Evolutionary Testing



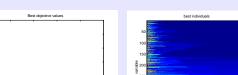
• each individual represents a test datum • execution time for each test datum

determines its objective value • when looking for worst-case execution time: test data with long execution time obtain high fitness values and vice versa employed tools:

- •GEATbx: Genetic and Evolutionary Algorithm Toolbox for Matlab (www.geatbx.com)
- QUANTIFY by Rational Software Corp.
- (www.rational.com)

5 Experiments

Optimization of bubblesort modul



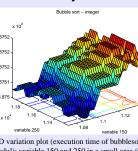
- execution time measured by **QUANTIFY** (Rational)
- Evolutionary Algorithm for integer/permutation variables: 6 subpopulations (regional model),

 different strategies per subpopulation

each 50 indiv., 1000 generations

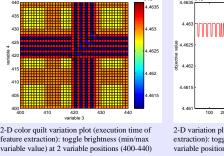
4 Search space analysis

- temporal behavior forms a very complex multi-dimensional search space
- many plateaus, local minima and discontinuities
 - several input data executing the same program path are identical
- different program paths lead to irregular changes of the exec. times two analysis examples provided
 - bubblesort modul (benchmark) feature extraction ME (real world)



Analysis of bubblesort modul

- •list of 500 variables, each variable in range [-32768, 32767]
- bubble sort variables • analysis results: many plateaus many local minima correlation between variables



Analysis of feature extraction modul (ME)

2-D variation plot (execution time of feature extraction): toggle 1 brightness value over

- 843 variables in [0, 4095]
- defining a 29x29 pixel matrix out of an
- 287x1200 pixel picture analysis results:
 - large steps in objective values even for small changes in variable values large plateaus without
 - change of objective value

6 Summary

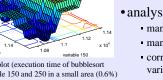
New approach:

Testing temporal behavior by Evolutionary Testing

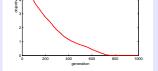
Advantages:

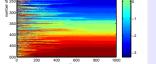
- + automatic search for the longest and shortest execution times
- + suited for discontinuous target functions

modul): variable 150 and 250 in a small area (0.6%) of the search space









Optimization of bubblesort modul (m execution time); best objective value over 1000 left); variables of best individual per generation generations, optimum for nd in generation 746

Optimization of bubb over 1000 generations ("sorting of variables") discrete recombination / order crossover integer mutation and swap mutation

competition between subpopulations

Optimization of motor control software modules

• results were compared to the times determined by the developers with systematic testing

	max. exec. time in µs			
module	evolutionary	developer	lines of	para-
name	testing	test	code	meters
zr2	69,6 µs	67,2 µs	41	18
t1	120,8 µs	108,4 µs	119	18
mc1	112,0 µs	108,4 µs	98	17
mr1	68,8 µs	64,0 µs	81	32
k1	59,6 µs	57,6 µs	39	14
zk1	58,4 µs	54,0 µs	56	9

• exec. time measured on target system • Evolutionary Algorithm for integer

- variables (parameter optimization):
 - 3 subpopulations (regional model), each 20 individuals over 100 generations
 - use of different strategies per subpopulation
 - discrete and double point recombination
- integer mutation (different mutation range) competition between subpopulations
- Evolutionary Testing found longer execution times for all given modules

- suited for complex input domains with many parameters +
- can escape local optima (search by multiple individuals) +
- direct assessment of objective value: execution time of individual +

But:

- finding the extreme execution times is not guaranteed
- Combination of systematic testing and Evolutionary Optimization opens up further potential for improvement
 - \rightarrow perform systematic test to examine functional correctness
 - \rightarrow use systematically produced test data to inoculate initial population of **Evolutionary Algorithm**
 - \rightarrow apply Evolutionary Testing to find extreme execution times

References

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