

## Automated Combinatorial Testing for Software

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### What is NIST? National Institute of Standards and Technology



- The nation's measurement and testing laboratory
- 3,000 scientists, engineers, and support staff including
  3 Nobel laureates
- Best known for atomic clock, standard reference materials (for instrument calibration) from aluminum alloy to whale blubber
- Basic and applied research in physics, chemistry, materials, electronics, computer science

### **Automated Combinatorial Testing**

 Project to combine automated test generation with combinatorial methods

• Goals – reduce testing cost, improve cost-benefit ratio for formal methods





### Overview of useful results

Proof of concept demo integrating combinatorial testing with model checking

• (Small) experimental result consistent with earlier interpretation of empirical data

 New combinatorial algorithms and tools, supporting development tradeoffs





### Problem: the usual ...

- Too much to test
- Even with formal specs, we still need to test
- Take advantage of formal specs to produce tests also – better business case for FM
- Testing may exceed 50% of development cost



- Example: 20 variables, 10 values each
- •10<sup>20</sup> combinations
- Which ones to test?



#### **Solution: Combinatorial Testing**

- Suppose no failure requires more than a pair of settings to trigger
- Then test all pairs 180 test cases sufficient to detect any failure





#### Pairwise testing – what do we know?

- Mandl, 1985 very effective for compiler test
- Brownlie, Prowse, Phadke high coverage
- Cohen, Dalal, Parelius, Patton, 1995 90% coverage with pairwise, all errors in small modules found
- Dalal, et al. 1999 effectiveness of pairwise testing, no higher degree interactions
- Smith, Feather, Muscetolla, 2000 88% and 50% of flaws for 2 subsystems,

What if finding ~90% of flaws is not good enough?



# How many combinations do we need to test to find ALL errors?

 Surprisingly, no one had looked at this question when NIST studied medical device software in 1999



- Wallace, Kuhn 2001 medical devices

   98% of flaws were pairwise interactions, no failure required > 4 conditions to trigger
- Kuhn, Reilly 2002 web server, browser; no failure required > 6 conditions to trigger
- Kuhn, Wallace, Gallo 2004 large NASA distributed database; no failure required > 4 conditions to trigger
- Max failure triggering fault interaction (FTFI) number of these applications was <u>6</u>
- Much more empirical work needed









#### FTFI numbers for 4 application domains – failures triggered by 1 to 6 conditions





#### Combinatorial test example: 5 parameters, 4 values each, 3-way combinations



## Problem: Combinatorial Testing Requires a Lot of Tests

- Number of tests: suppose we want all 4-way combinations of 30 parameters, 5 values each: 3,800 tests – too many to create manually
- Test set to do this is a <u>covering array</u>
- Time to generate covering arrays: problem is NP hard
- No. of combinations:  $\binom{n}{k} v^k$



For *n* variables with *v* values, *k*-way combinations



## **Solution: Automated Testing**

Test data generation – easy

- Test oracle generation hard
- Creating test oracles model checking and other state exploration methods
- Model-checker test production: if assertion is not true, then a counterexample is generated. This can be converted to a test case.





#### Using model checking to produce tests



#### Model checking example



```
-- specification for a portion of tcas - altitude separation.
-- The corresponding C code is originally from Siemens Corp. Research
-- Vadim Okun 02/2002
MODULE main
VAR
  Cur Vertical Sep : { 299, 300, 601 };
  High Confidence : boolean;
. . .
init(alt sep) := START ;
  next(alt sep) := case
    enabled & (intent not known | !tcas equipped) : case
      need upward RA & need downward RA : UNRESOLVED;
      need upward RA : UPWARD RA;
      need downward RA : DOWNWARD RA;
      1 : UNRESOLVED;
    esac;
    1 : UNRESOLVED;
  esac;
. . .
SPEC AG ((enabled & (intent not known | !tcas equipped) &
!need downward RA & need upward RA) -> AX (alt sep = UPWARD RA))
```



#### **Computation Tree Logic**

```
The usual logic operators, plus temporal:
  A \varphi - All: \varphi holds on all paths starting from the current state.
  E \varphi - Exists: \varphi holds on some paths starting from the current state.
  G \varphi - Globally: \varphi has to hold on the entire subsequent path.
  F \varphi - Finally: \varphi eventually has to hold
  X \varphi - Next: \varphi has to hold at the next state
      [others not listed]
     execution paths
            states on the execution paths
SPEC AG ((enabled & (intent not known |
!tcas equipped) & !need downward RA &
need upward RA)
\rightarrow AX (alt sep = UPWARD RA))
```

```
"FOR ALL executions,
IF enabled & (intent_not_known ....
THEN in the next state alt_sep = UPWARD_RA"
```

## How can we integrate combinatorial testing with model checking?

- Given AG (P -> AX (R)) "for all paths, in every state, if P then in the next state, R holds"
- For k-way variable combinations, v1 & v2 & ... & vk
- vi abbreviates "var1 = val1"
- Now combine this constraint with assertion to produce counterexamples. Some possibilities:
  - AG(v1 & v2 & ... & vk & P -> AX !(R))
  - $AG(v1 \& v2 \& \ldots \& vk \rightarrow AX ! (1))$
  - AG(v1 & v2 & ... & vk -> AX !(R))



What happens with these assertions?

1. AG(v1 & v2 & ... & vk & P -> AX !(R))

P may have a negation of one of the  $v_i$ , so we get 0 -> AX ! (R))

always true, so no counterexample, no test. This is too restrictive

1. AG (v1 & v2 & ... & vk  $\rightarrow$  AX ! (1))

The model checker makes non-deterministic choices for variables not in v1..vk, so all R values may not be covered by a counterexample. This is too loose

2. AG (v1 & v2 & ... & vk  $\rightarrow$  AX !(R))

Forces production of a counterexample for each R. This is just right





#### Proof-of-concept experiment

- Traffic Collision Avoidance System module
  - Small, practical example 2 pages of SMV
  - Used in other experiments on testing
    - Siemens testing experiments, Okun dissertation
  - Suitable for model checking
- 12 variables: 7 boolean, two 3-value, one 4-value, two 10-value
- Tests generated w/ Lei "In Parameter Order" (IPO) algorithm extended for >2 parameters



#### **Combinations /tests generated**

t	Comb.	Test cases
2-way:	100	156
3-way:	405	461
4-way:	1,375	1,450
5-way:	4,220	4,309
6-way:	10,902	11,094

(more "don't care" conditions at lower interaction levels)



#### **Results**



- Roughly consistent with data on large systems
- But errors harder to detect than real-world examples





# What do we need to make this practical?

- This approach would not have been practical 10 years ago
- Now we have high performance model checkers, better covering array algorithms, and cheap processors
- Generating ~  $10^6 10^7$  tests can be done
- Proof of concept experiment completed

So what? Finding covering arrays is an NP hard problem!

#### Solution: new covering array algorithms

- Tradeoffs to minimize calendar/staff time:
- FireEye (extended IPO) Lei roughly optimal, can be used for most cases under 40 or 50 parameters
  - Produces minimal number of tests at cost of long run time
  - Currently integrating algebraic methods
- Adaptive distance-based strategies Bryce dispensing one test at a time w/ metrics to increase probability of finding flaws
  - Highly optimized covering array algorithm
  - Variety of distance metrics for selecting next test
- Paintball Kuhn for more variables or larger domains
  - Randomized algorithm, generates tests w/ a few tunable parameters; computation can be distributed
  - Better results than other algorithms for larger problems





## Will automated combinatorial testing work in practice?

The usual potential pitfalls:

. Faithfulness of model to actual code

.Always a problem

•Being able to generate tests from specification helps make formal modeling more cost effective

- . Time cost of generating tests
  - Model checking very costly in run time

• Inherent limits on number of variable values even with ideal covering array generation: need at least  $C(n,k) * v^k$ 

- Abstraction needed to make this tractable
  - Equivalence classes for variable values may miss a lot that matters

 Not all software is suited to this scheme – e.g., good for code with lots of decisions, not so good for numerical functions.





- Two real-world trials planned US Govt Personal Identity Verification (PIV) card, machine tool specification exchange software
- Plan to experiment with both SMV model checker and TVEC
- Generate 10<sup>5</sup> to 10<sup>6</sup> tests per module, probably up to 5way combinations





#### Summary and conclusions

- Proof of concept is promising integrated w/ model checking
- Appears to be economically practical
- New covering array algorithms help make it more tractable
- Cluster implementation of covering array algorithm
- Many unanswered questions
  - Is it cost-effective?
  - What kinds of software does it work best on?
  - What kinds of errors does it miss?
  - What failure-triggering fault interaction level testing is required? 5way? 6-way? more?
- Large real-world example will help answer these questions

Please contact us if you are interested!

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