

Software Verification / Testing

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This Talk

Some recent developments in software
verification and testing

Software Verification?

- Related to, but different from, IEEE definition
- Traditionally, in CS: *formal methods*
 - Given software, spec
 - Software = “code”
 - Spec = “requirement” = logical formula
 - Prove software meets spec
- (Informal verification often called “validation”.)

Model Checking

- Verification = proof
- Model checking: automated proof!
 - Given software, spec
 - Model checker tries to build proof
- Ongoing research: applicability
 - Decidability
 - Scalability
- Embedded control applications!

Software Testing

- Most often-used method for checking software correctness
 - Select tests
 - Run software on tests
 - Analyze results
- Traditionally
 - Manual, hence time-consuming, expensive
 - In control applications: hard to test software by itself

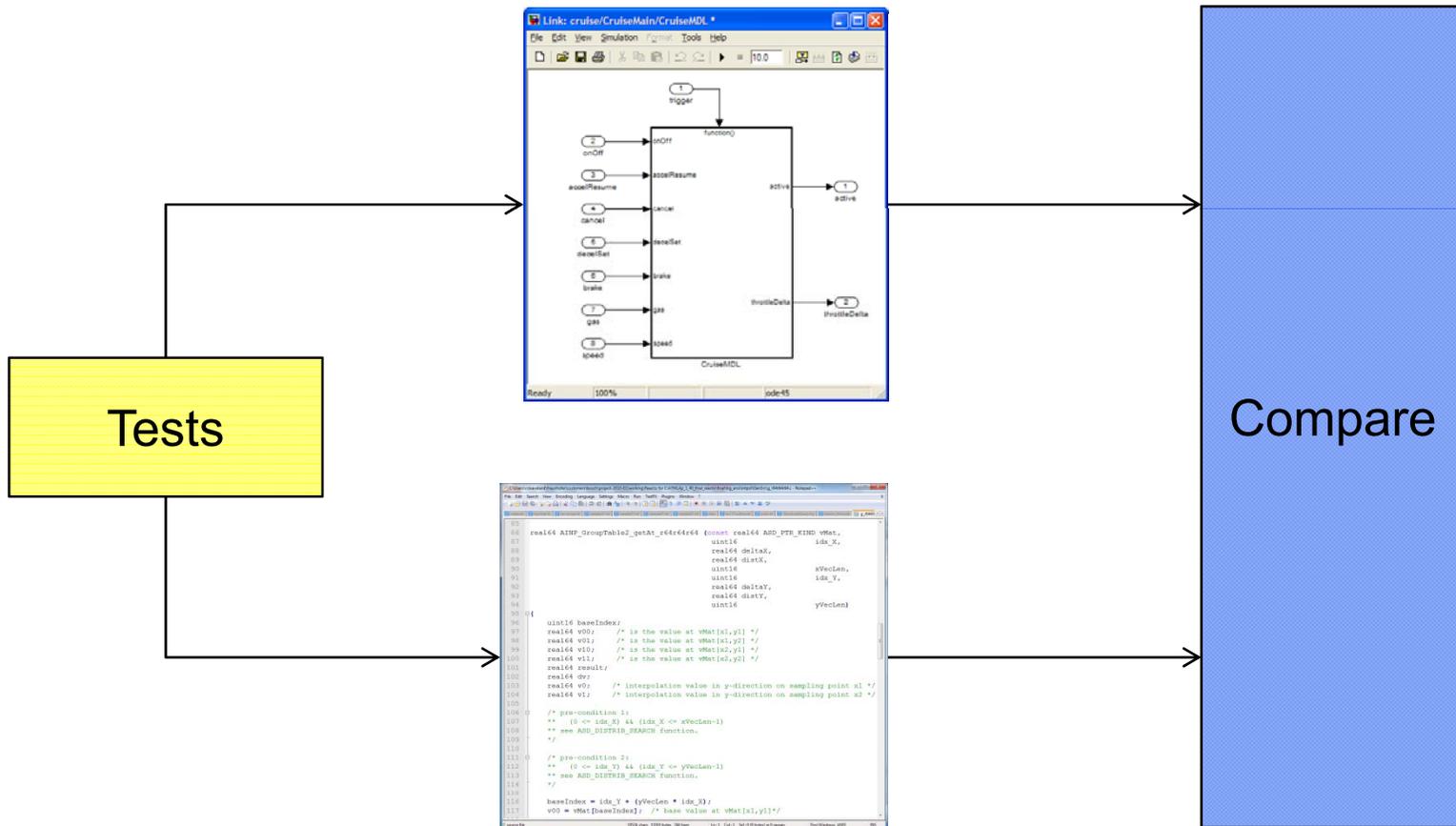
Exciting Developments

- Combine
 - Formal specs
 - Testing
- To automate testing “scalably”
 - Model-based testing
 - Instrumentation-based verification
 - Requirements reconstruction

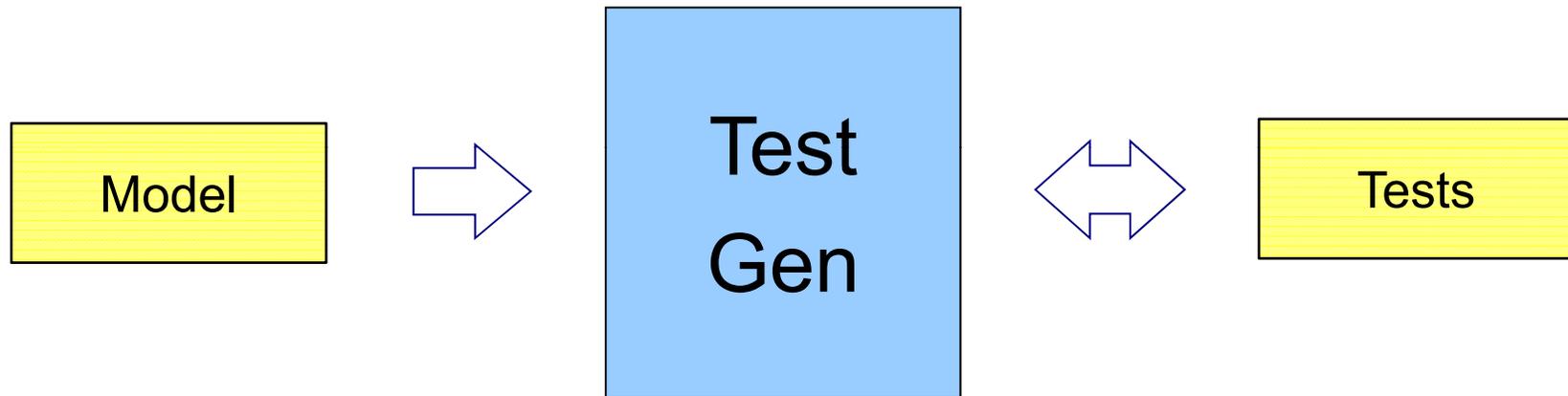
Model-Based Testing

- Develop specs as executable models
 - Simulink
 - State machines
 - Etc.
- Use model to determine correct test response
 - Automates “results analysis”
 - Models, tests needed

Model-Based Testing (cont.)



Tests Can Be Generated from Models!



- Functionality provided by tools like Reactis® for Simulink / Stateflow
- Goal: automate test generation task by creating tests that cover model logic
- Reactis: guided simulation algorithm

Applying Model-Based Testing

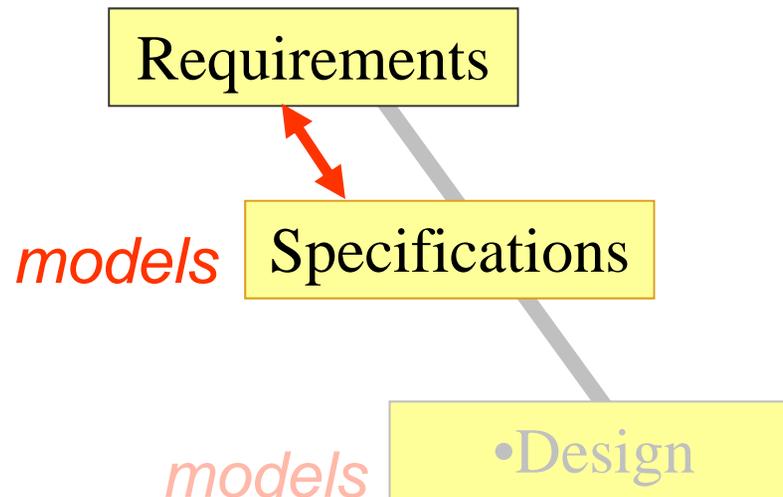
- Widespread in automotive, less so in aero / medical-device
 - Regulatory issues
 - Need for models
 - Modeling notations, support
- What about models?
 - Sometimes result of earlier design phases
 - Models as reusable testing infrastructure

Challenges

- Technical
 - Algorithms for test generation
 - Modeling languages
- Procedural
 - Integration into existing QA processes
 - Regulatory considerations

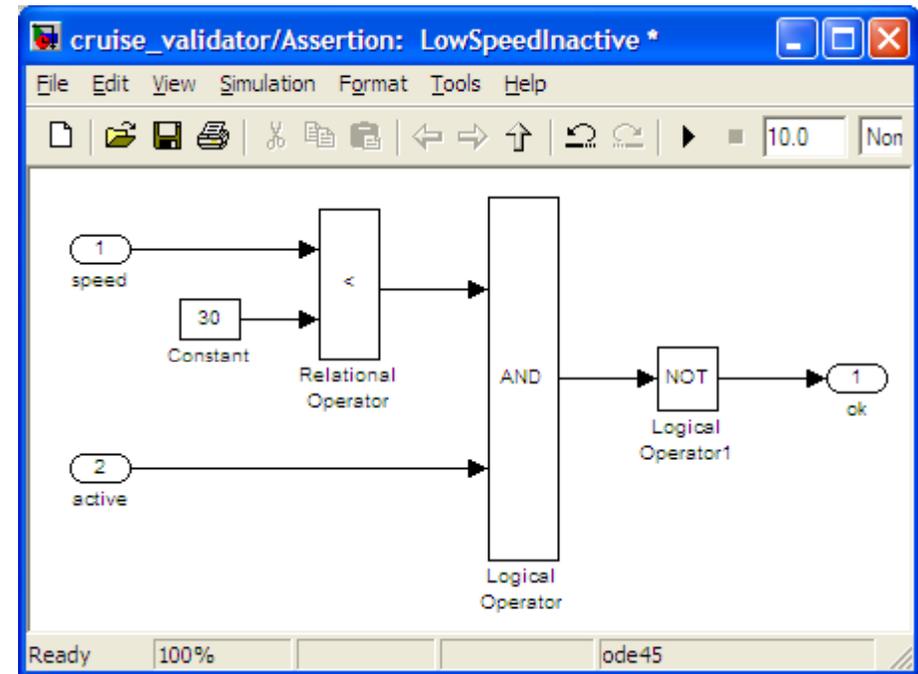
Instrumentation-Based Verification

- Model-based testing assumes model correct
- IBV: a way to check model correctness vis a vis requirements



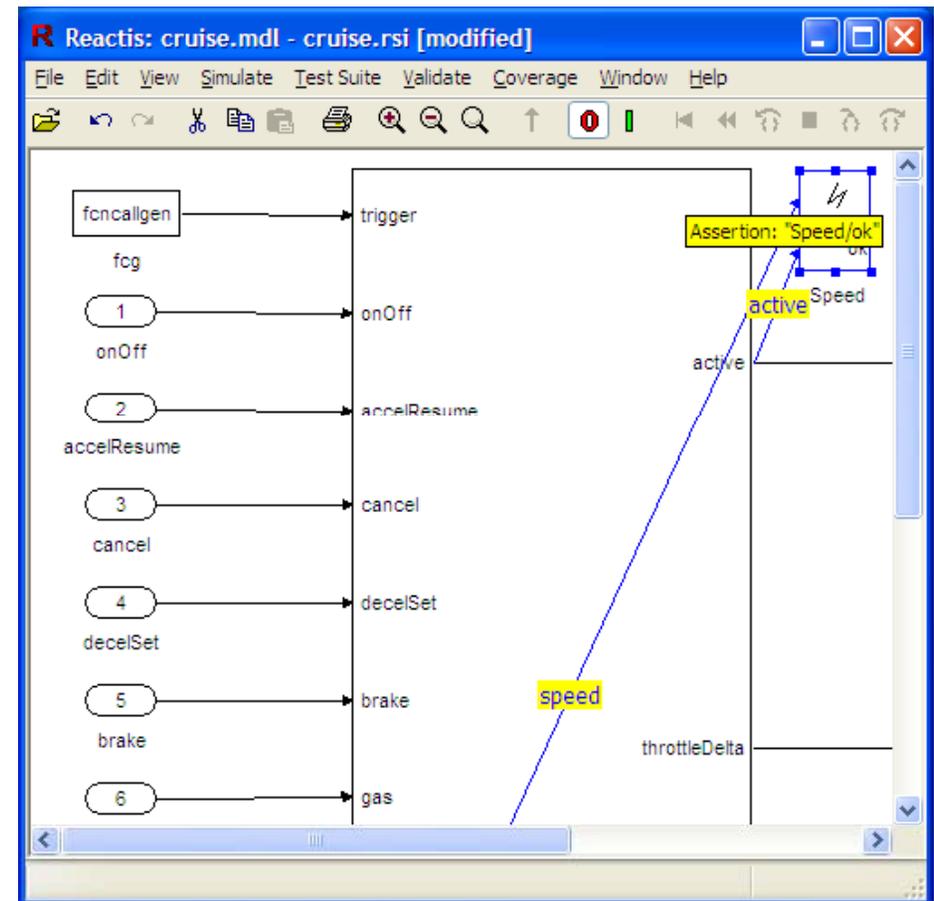
Instrumentation-Based Verification: Requirements

- Verification needs formalized requirements
- IBV: formalize requirements as *monitor models*
- Example
“If speed is < 30 , cruise control must remain inactive”



Instrumentation-Based Verification: Checking Requirements

- Instrument design model with monitors
- Use *coverage testing* to check for monitor violations
- Tool: Reactis®
 - Product of Reactive Systems, Inc.
 - Separates instrumentation, design
 - More info: www.reactive-systems.com



Applying Instrumentation-Based Testing

- Robert Bosch production automotive application
 - Requirements: 300-page document
 - 10 subsystems formalized (20% of system)
 - 62 requirements formalized as monitor models
 - IBV applied
 - 11 requirements issues identified
- Another Bosch case study: product-line verification using IBV
- A number of other case studies

Requirements Reconstruction

- The Requirements Reconstruction problem
 - Given: software
 - Produce: requirements
- Why?
 - System comprehension
 - Specification reconstruction
 - Missing / incomplete / out-of-date documentation
 - “Implicit requirements” (introduced by developers)

Invariants as Requirements

- Some requirements given as invariants
 - *“When the brake pedal is depressed, the cruise control must disengage”*
- State machines can be viewed as invariants
 - States: values of variables
 - Transitions: invariants
 - *“If the current state is A then the next state can be B”*
- Another project with Robert Bosch

Invariant Reconstruction

- Generate test data satisfying *coverage criteria*
- Use *machine learning* to propose invariants
- Check invariants using *instrumentation-based verification*

Machine Learning: *Association Rule Mining*

- Tools for inferring relationships among variables based on time-series data

– Input: table

Time	x	y
0	1	0
1	-1	-1
2	2	1
...

– Output: relationships (“association rules”)

e.g. $0 \leq x \leq 3 \rightarrow y \geq 0$

Association Rules and Invariant Reconstruction

- General idea
 - Treat tests (I/O sequences) as data
 - Use machine learning to infer relationships between inputs, outputs
- Our insight
 - Ensure test cases satisfy coverage criteria to ensure “thoroughness”
 - Use IBV to double-check proposed relationships

Pilot Study: Production Automotive Application

- Artifacts
 - Simulink model (ca. 75 blocks)
 - Requirements formulated as state machine
 - Requirements correspond to 42 invariants defining transition relation
- Goal: Compare our approach, random testing [Raz]
 - Completeness (% of 42 detected?)
 - Accuracy (% false positives?)

Experimental Results

- Hypothesis: coverage-testing yields better invariants than random testing
- Coverage results:
 - 95% of inferred invariants true
 - 97% of requirements inferred
 - Two missing requirements detected
- Random results:
 - 55% of inferred invariants true
 - 40% of requirements inferred
- Hypothesis confirmed

Summary

- Intersection of formal methods, testing can yield practical verification approaches
 - Model-based testing
 - Instrumentation-based verification
- Automated test generation can be used to infer invariants