

Formal Methods Meets Testing

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The
Institute for
Systems
Research

This Talk: PIYC / TIYC



- “Pick-tick”
 - Prove If You Can.
 - Test If You Cannot.
- More precisely
 - Formal specifications should support verification
 - They should also support testing
 - Testing should be seen as “approximate formal verification”

Formal Methods



- Mathematically rigorous approaches to specifying, verifying systems
 - Originally: software, hardware design
 - Key people: Clarke, Dijkstra, Hoare, Lamport, Milner, Pnueli, ...
- Why? To increase confidence!
 - If the specification is trusted, verification yields trust in system
 - If specification is not trusted, proving it is consistent with system builds trust in both

The Elements of Formal Methods



- Formal semantics of systems (e.g. state machines)

Systems must be mathematical objects!

- Formal specifications (e.g. temporal logic)

Mathematical descriptions of desired behavior

- Formal verification = proof

- Model checking: **Proofs done automatically**

- Theorem proving: **Proofs done “automatedly”**

Status of Formal Methods



- Noteworthy successes!
 - sel4 OS kernel verification
 - Railway signaling
 - Paris Roissy VAL shuttle
 - Mars Rover
 - Satellite control
 - ...
- We are not at the stage where success is expected

Why?



- “Scalability”

Building proofs is laborious, even for machines

- Inability to predict level of effort

- Difficulty of proof not correlated to usual measures of system complexity
- Work needed to coax proof out of tools not easy to estimate

- How verification and validation happens in practice
 - Incomplete, but
 - Scalable, and
 - Mandated (regulation)
- Terminology
 - Black box / white box
 - Hardware-in-the-loop
 - Model-based testing (MBT)
 - ...

My Perspective



- Proving is hard, but guarantees are very strong
- If proofs are not possible
 - Must test to conduct V&V
 - Benefits of formal specifications in this case?
- “Prove If You Can, Test If You Cannot”
(PIYC/TIYC)

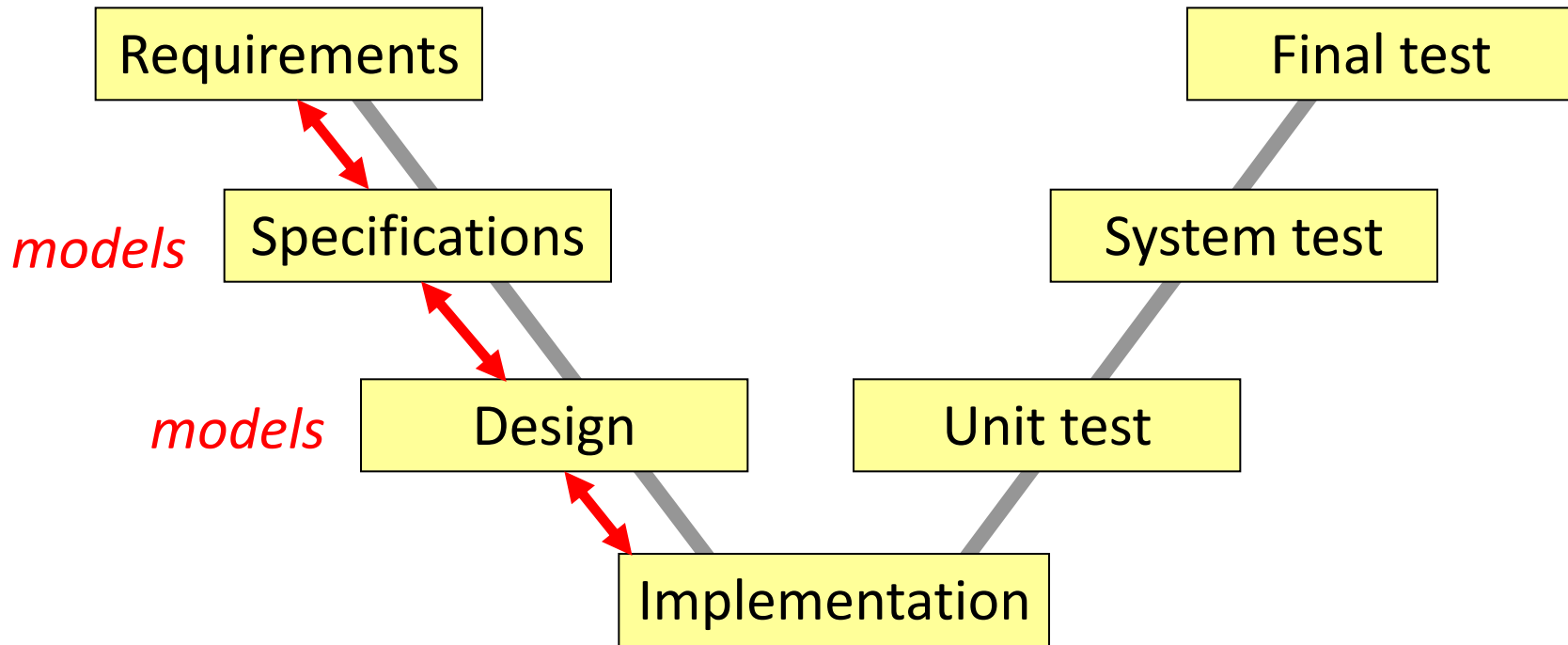
We should focus on formal specifications that support proof *and* testing!

Rest of Talk: PIYC / TIYC in Practice



- Model-based testing
 - Models used as software specifications
 - MBT used to check software *vis à vis* specs
- Instrumentation-Based Verification (IBV)
 - Specifications given in same notation as software
 - Verification = instrument software, check for errors
- Context
 - Automotive control software and Model-Based Development (MBD)
 - MATLAB[®] / Simulink[®] / Stateflow[®] / Reactis[®]

Model-Based Development (MBD) in Automotive



Main Motivation: *autocode*

Models also support *V&V, testing*

PIYC / TIYC for MBD

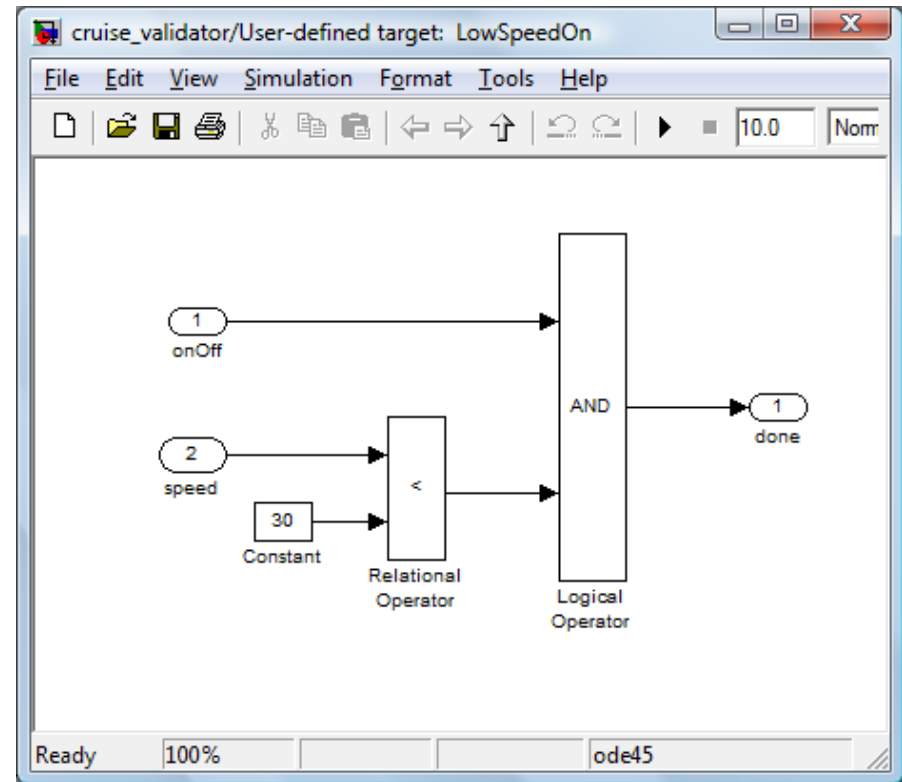


- Formalize verification problems mathematically
- Give testing-based *approximate* verification strategies based on formalizations

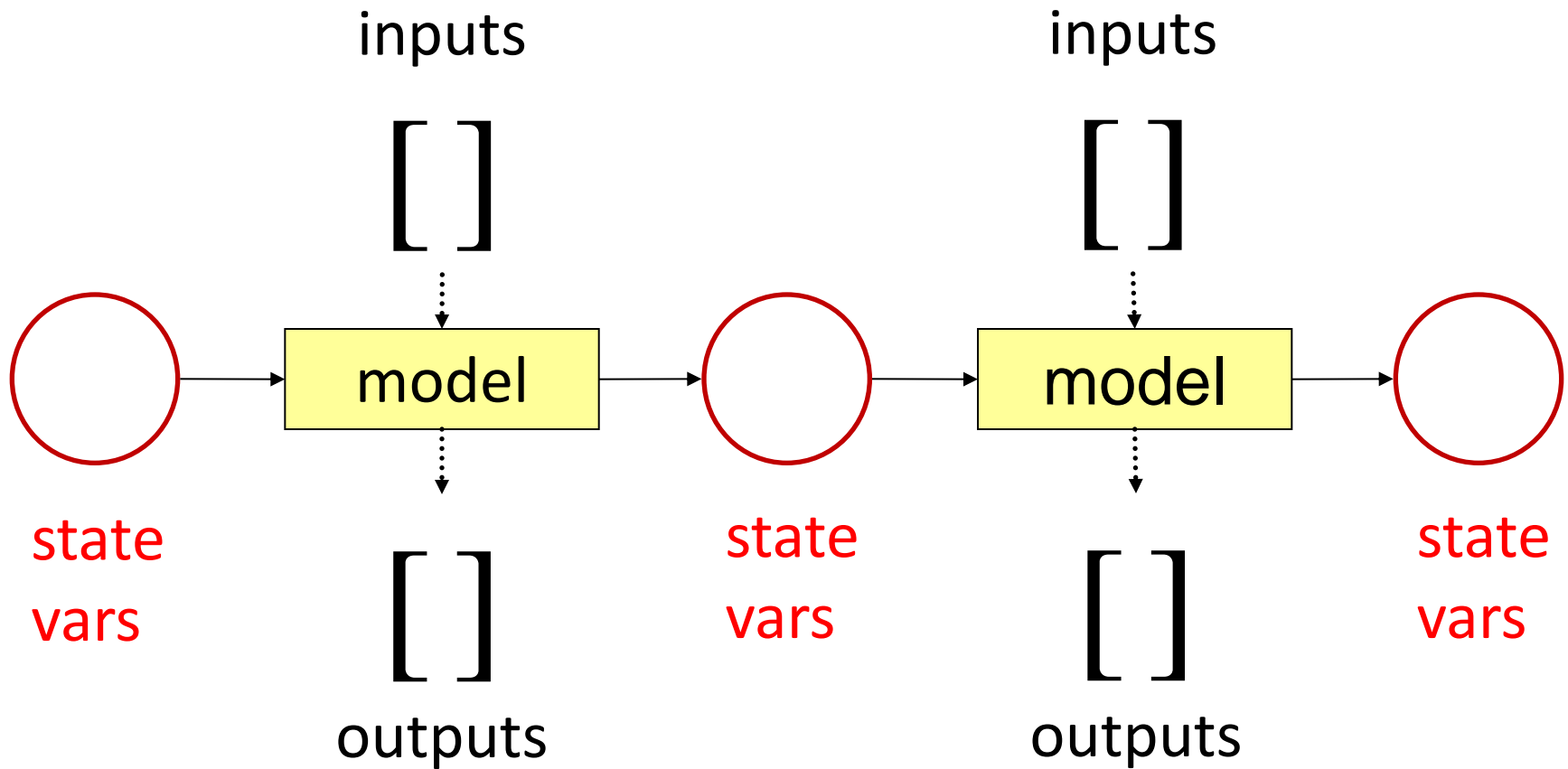
Simulink



- Block-diagram modeling language / simulator of The MathWorks, Inc.
- Hierarchical modeling
- Continuous- and discrete-time simulation



Discrete Simulink Semantics

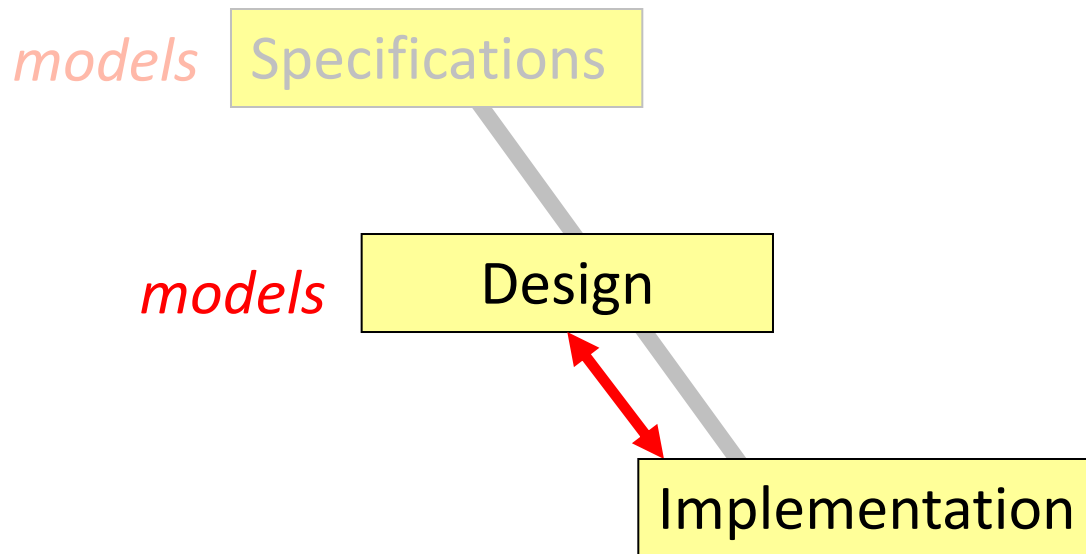


Discrete Simulink Semantics (cont.)



- Simulink models are *Mealy machines*
 - States are assignments to state variables
 - Transitions are computed by model
- Can thus speak of *language* of model M
 - I = set of possible input vectors for M
 - O = set of possible output vectors for M
 - $L(M) = \{w \in (I \times O)^* \mid w \text{ is sequence of transition labels of execution of } M \}$

MBD Verification Problem #1

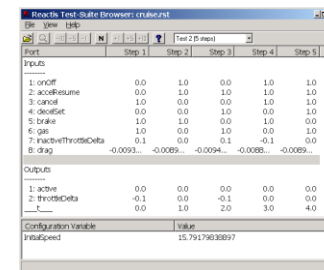
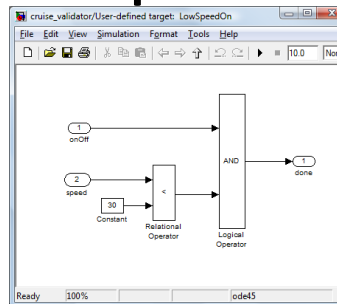


Does implementation meet design?

Model-Based Testing

- An emerging approach to this problem

- From Simulink model ...
- ... generate test cases
- ... and run them on system
- ... comparing results



Step	Step 1	Step 2	Step 3	Step 4	Step 5
1: onOff	0.0	1.0	0.0	1.0	1.0
2: accelResume	0.0	1.0	1.0	1.0	1.0
3: cancel	1.0	0.0	0.0	1.0	1.0
4: onOffSet	0.0	0.0	1.0	0.0	1.0
5: brake	1.0	1.0	0.0	1.0	0.0
6: gas	1.0	0.0	1.0	0.0	1.0
7: inactiveThrottleDelta	0.1	0.0	-0.1	-0.1	0.0
8: drag	-0.0093...	-0.0093...	-0.0094...	-0.0085...	-0.0089...
1: active	0.0	0.0	0.0	0.0	0.0
2: throttleDelta	-0.1	0.0	-0.1	0.0	0.0
3: ...	0.0	1.0	2.0	3.0	4.0

Configuration Variable Value
InitialSpeed 15.79179038897

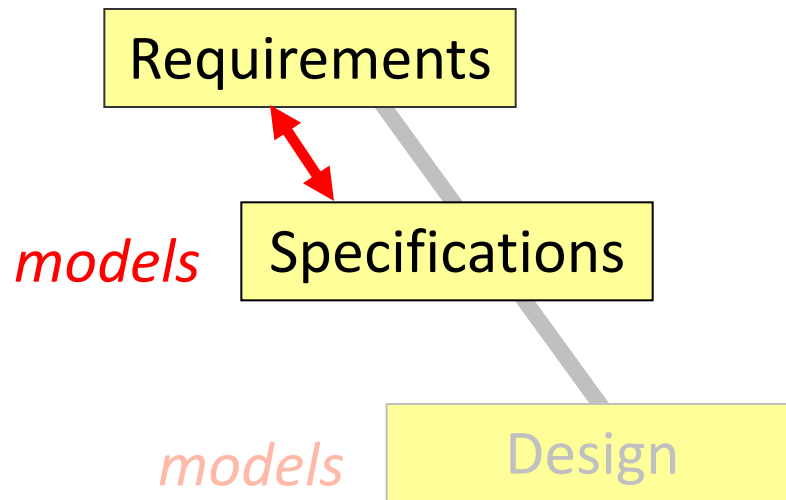
- Model serves as
 - Specification
 - Test oracle

Improving MBT



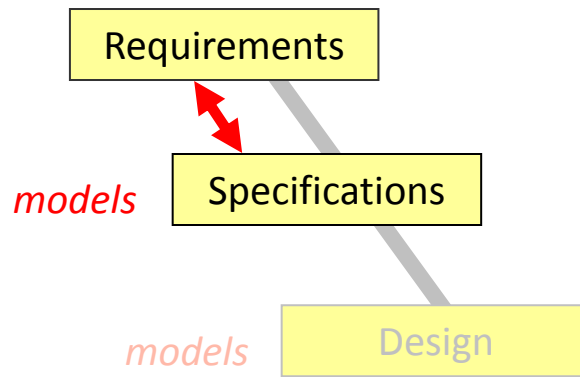
- Recall formalization of Problem #1
 - Given M, S
 - Prove $L(M) = L(S)$
 - Classical MBT: generate tests from M , run on S
- ***If $L(M) = L(S)$ is goal, why not also generate tests from S , run them on M ?***
- Result: “back-to-back testing”
 - Reactis used to generate tests from M
 - Reactis for C used to generate tests from C code
 - Controversial!
 - “You can’t generate tests from implementations”
 - But formalization suggests this is perfectly reasonable!

MBD Verification Problem #2



Do specs satisfy requirements?

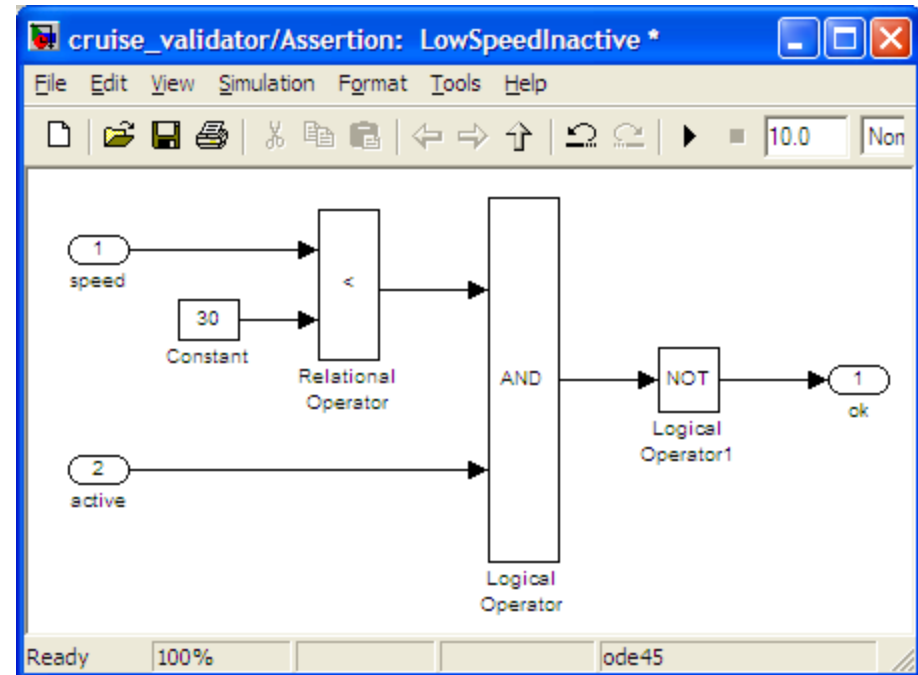
Formalizing Verification Problem #2



- Need following for PIYC / TIYC:
 - Formalized requirements
 - Formalized notion of satisfaction
- Our approach: *Instrumentation-Based Verification*

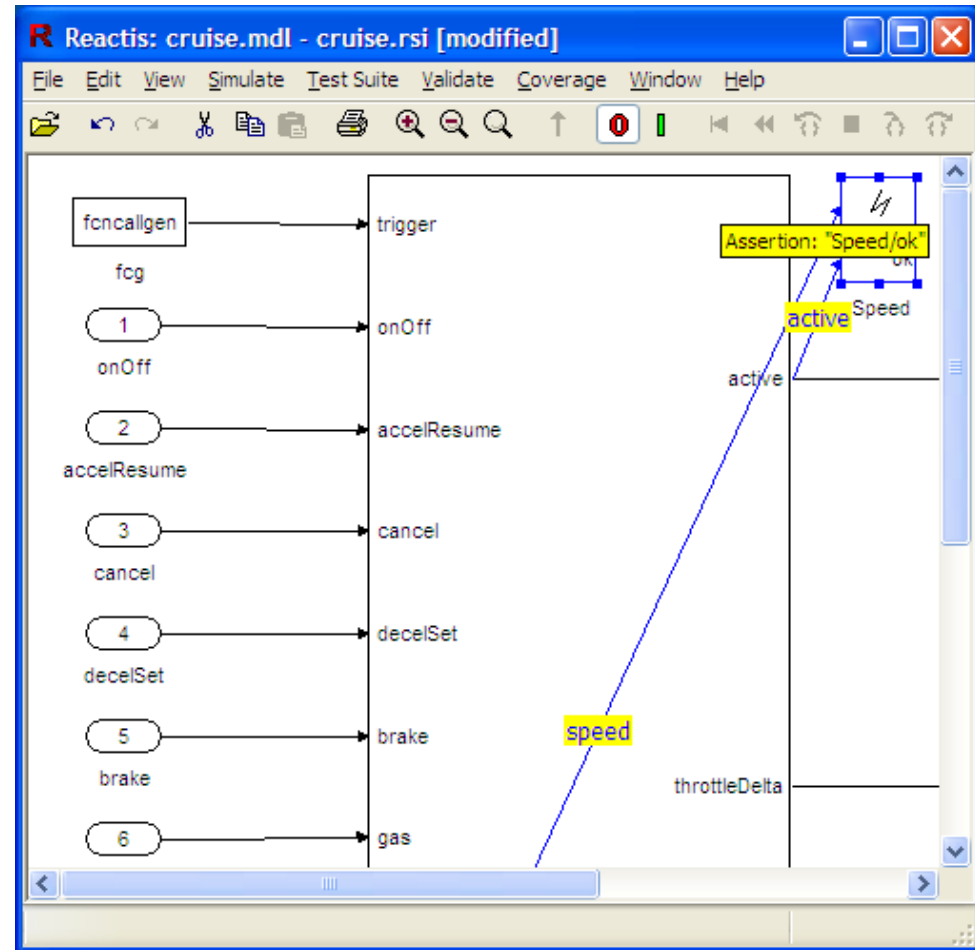
IBV: Requirements

- Formalize requirements as *monitor models*
- Example
If speed is < 30, cruise control must remain inactive



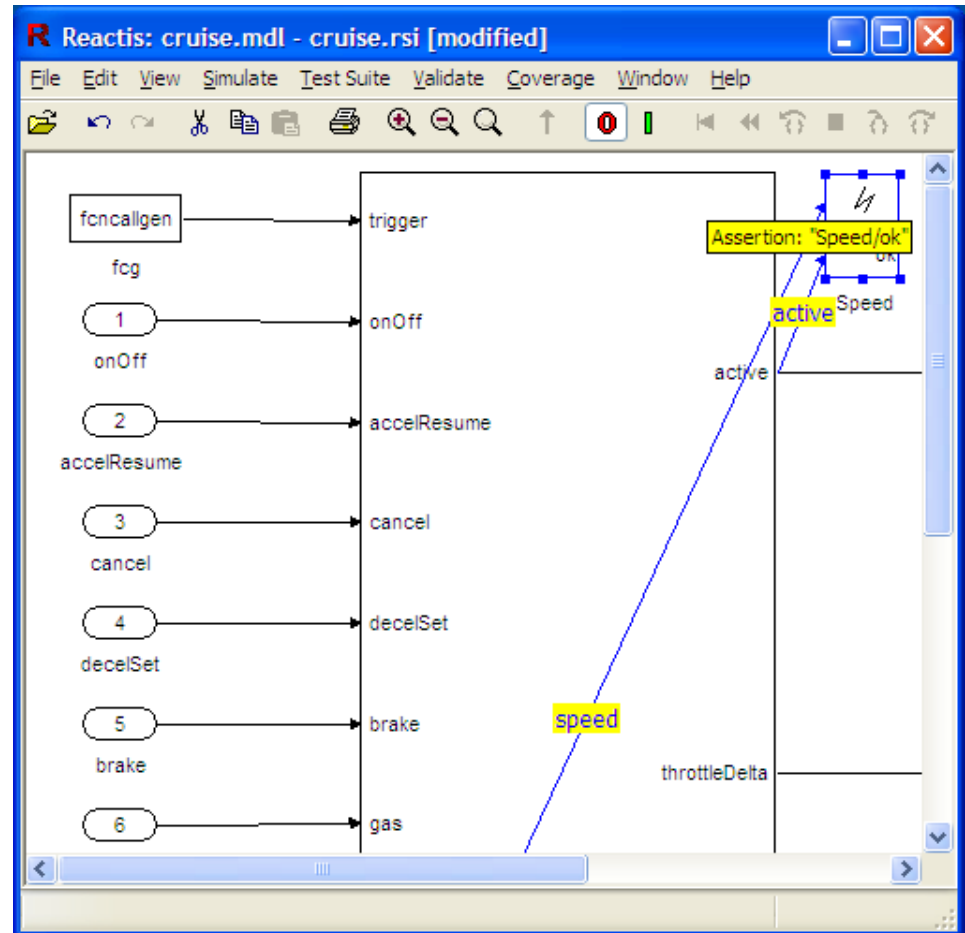
IBV: Satisfaction

- Instrument design model with monitors
- Model satisfies monitors if:
 - For every input sequence ...
 - ... every monitor model output remains *true*
- Reachability problem!
 - Proof possible
 - State space an issue



Approximate Verification for Problem #2

- Use coverage testing on instrumented model
 - Better scalability
 - If booleans part of coverage criteria:
 - Test generator tries to make monitor outputs false
 - **Skeptical testing!**
- Reactis
 - Supports instrumentation
 - Acts as skeptical tester
 - Reports violations



Summary



- PIYC / TIYC!
 - Formalize specs
 - View testing as “approximate” formal verification
- Applications in model-based testing, verification against requirements

Provocative Statement!



- Down with Temporal Logic!
- Really?
 - Of course not! Great vehicle for research
 - But PIYC/TIYC? Not so much ...

Specification Reconstruction



- V&V needs requirements specifications
 - Requirements then checked using testing, formal methods, etc.
 - Quality of V&V depends on quality of specification
- Problem!
 - Specification must be maintained, updated, checked
 - Implicit requirements often not documented
 - “Emergent behavior?”
- *Specification reconstruction*
 - Given system (model) ...
 - ... automatically propose requirements

White-Box Invariant Reconstruction



- Invariants: one type of requirements
 - Invariant stipulates relationship that should be preserved among state variables as system evolves
 - E.g.
 - If the brake pedal is pressed, the cruise control must immediately disengage*
- Invariant reconstruction via models, data mining, IBV
 - Generate test cases
 - Compute proposed invariants using data mining
 - Check proposed invariants using IBV
 - Repeat

(Joint work with Christoph Schulze)

- Tools for inferring patterns in (time-series) data

– Input: table

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

– Output: patterns (= formulas)

e.g. $-1 \leq x \leq 2$

$0 \leq x \leq 3 \rightarrow y \geq 0$

Association Rules



- An important class of patterns!
 - Form: $\bigwedge \phi_i \rightarrow \bigwedge \psi_j$
 - ϕ_i, ψ_j are propositions involving variables, constants
 - $\{\phi_i\}, \{\psi_j\}$ are disjoint
 - Traditionally: $j = 1$
 - E.g. $x = 1 \rightarrow y = 0$
- Our work: find invariants in form of association rules
 - LHS: propositions involving inputs, “incoming state”
 - RHS: proposition involving outputs, “outgoing state”

Apriori Algorithm



- Widely used association-rule mining technique
- Developed in 1993-94 by Agrawal et al.
 - SIGMOD 1993
 - VLDB 1994
- Implemented in many data-mining tools (Weka, Magnum Opus, ...)

Invariant Reconstruction



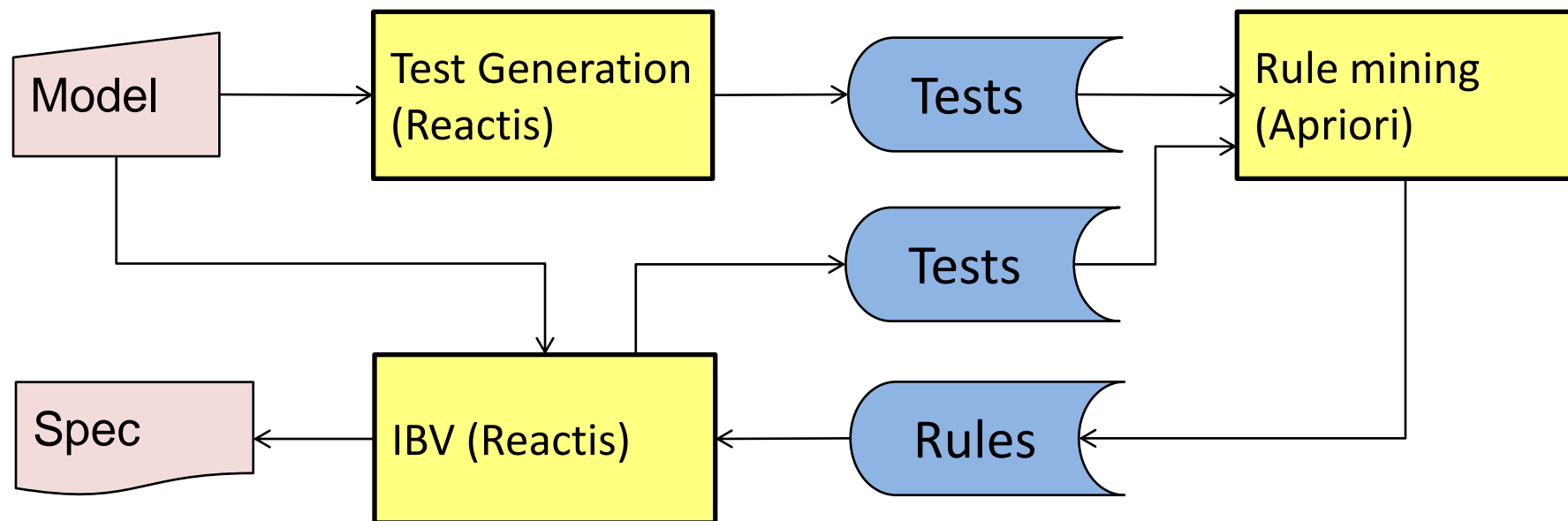
- General idea
 - Treat test results (I/O sequences) as “tables”
 - Invariants: association rules with coverage ≥ 1 , strength = 1
 - Use Apriori to compute invariants involving inputs (antecedent), outputs (consequent)
- Additionally
 - Ensure test cases satisfy structural coverage criteria (e.g. branch coverage) to ensure “thoroughness”
 - Use IBV to double-check proposed invariants

What About IBV Tests?



- In IBV, coverage testing of instrumented model used to check for monitor violations
- Tests inducing violations can be used to remove invariants subsequent “minings”
- They also can be a source of other invariants

Our Approach in Detail



- Reactis creates tests to do IBV check
- These tests are “cycled back through” the data-mining tool, together with original tests

Pilot Study: Production Body-Electronic Application



- Artifacts
 - Simulink model (ca. 75 blocks)
 - Requirements spec formulated as state machine
 - Requirements correspond to 42 invariants defining transition relation, e.g.
$$\text{state} = 1 \wedge \text{pressed} = \text{true} \rightarrow \text{new_state} = 2$$
- Goal: Compare our approach, random testing
 - Completeness (% of 42 detected?)
 - Accuracy (% false positives?)

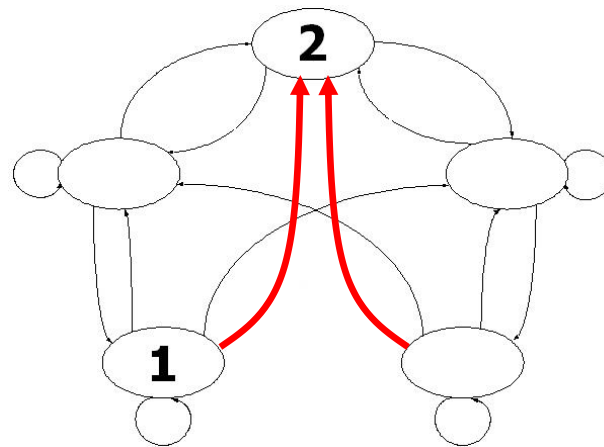
Experimental Results



- Hypothesis: coverage-testing yields better invariants than random testing
- Coverage results (one iteration of test generation)
 - 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 (for this case study)

Requirement Issue

- Missing reset transitions in requirements
- Code was correct



Procedural Issues



- How do you trust generated invariants in absence of “requirements baseline”?
- Our approach: *Jaccard similarity*

Jaccard Similarity Measures



- Jaccard: a tool for measuring set similarity
- Let A, B be sets. Then the **Jaccard similarity measure**, $J(A, B)$, of A and B is:

$$J(A, B) = \frac{|A \cap B|}{|A \cup B|}$$

- Facts
 - $0 \leq J(A, B) \leq 1$
 - Closer to 1 means “more similar”

Jaccard and Invariant Generation



- High Jaccard similarity means more “stable”
 - For coverage: average Jaccard score is 0.87
 - For random: average is 0.65
- Another use: iteration termination
 - Our approach allows iteration of “test / generate / check”
 - When to terminate: use Jaccard! (i.e. terminate when successive invariant sets are “similar enough”)



Provocative Statement!



Focus at UMD in CyberCardia



- Foundations, tools for reasoning about CPS
 - Formal modeling of CPS
 - Formal specification, verification
- This year: Specification reconstruction
 - Given model M , infer temporal properties that M (likely) satisfies
 - Motivations
 - Model understanding
 - Specification updating
 - Means for “jump-starting” formal specifications in often unfamiliar notations
- See poster (48-50)!

Specific Results in 2017



- Linear temporal-logic query checking
 - Problem
 - Given Kripke structure M , LTL “template” $\phi[x]$
 - Find most general solution ϕ' for missing formula x so that M satisfies $\phi[x:=\phi']$
 - Algorithmic solution based on model checking developed, implemented, evaluated
 - Work presented at AVoCS/FMICS 2017
- Invariant mining from test data
 - Problem
 - Given (Simulink) model M , state variables of interest
 - Propose invariants describing relationships among variables
 - Approach: use data-mining on test data coupled with retesting to generate likely invariants
 - Evaluation used 11 models from automotive, medical-device domain
 - Work presented at EMSOFT 2017