Advanced Verification and Validation Methods for Cyber-Physical Systems

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Outline

- 1. Cyber-physical Systems for Verification & Validation
 - Perspective
 - Analysis for CPS
 - Formal setting
 - Emerging techniques
- 2. Requirements Engineering
 - Ongoing challenges
 - ST-Lib: Library of formal requirements for CPS applications



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CPS will be everywhere!



http://www.toyota-global.com/innovation/intelligent_transport_systems/mobility/images/its_mobility_02_large.jpg

CPS is safety critical!

- CPSs used in safety critical applications
 - Automotive powertrain control
 - Smart grids
 - Aerospace control
 - Medical devices



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CPS is safety critical!







The FDA has issued 23 recalls of defective devices during the first half of 2010, all of which are categorized as "Class I," meaning there is "reasonable probability that use of these products will cause serious adverse health consequences or death."



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FORMAL ANALYSIS SETTING



Definition (System):

System \mathcal{M} is some manifestation of a dynamical system whose behaviors $\phi(\mathcal{M}, p, u)$ are determined by parameters p and inputs u

- Generally, \mathcal{M} can be a model, a test experiment (e.g., HILs, SILs), or the physical system
- For simulation and analysis, we will assume \mathcal{M} is a model of the system (e.g., Simulink)
- Note that p and u can be taken from (possibly infinite) sets P and U

$$u \in U$$

Parameters: $p \in P$
 $\mathcal{M} \longrightarrow \phi(\mathcal{M}, p, u)$



Definition (Simulation):

Process of generating $\phi(\mathcal{M}, p, u)$, which are the behaviors of \mathcal{M} given parameters p and inputs u

• Assume simulations can be generated by numerical integration solver (e.g., Simulink)



Definition (Testing):

Determine whether $\phi(\mathcal{M}, \hat{P}, \hat{U}) \vDash \varphi$ for given finite sets $\hat{P} \subseteq P$ and $\hat{U} \subseteq U$

• Testing does not guarantee φ holds for all $p \in P$ and $u \in U$



<u>Definition (Verification)</u>: Prove $\phi(\mathcal{M}, P, U) \vDash \varphi$ given P and U

• Proves φ holds for all $p \in P$ and $u \in U$



Software vs. Control Design

- Classical software design
 - Nontrivial verification questions for finite state models of software are hard
 - In general, proving nontrivial properties for software is undecidable
 - Σ_1 undecidable

- Embedded control system design
 - Nontrivial verification questions for even simple CPSs are very undecidable
 - Σ_2 undecidable⁺







• E. Asarin and O. Maler. Achilles and the Tortoise Climbing Up the Arithmetical Hierarchy. *Journal of Computer and System Sciences*, Vol. 57, No. 3, pp. 389-398, 1998.

Spectrum of Analysis Techniques



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SIMULATION-BASED CHECKS FOR POWERTRAIN CONTROL

Why simulations?

- Help design validation
- Provide visual feedback
- Can use existing design artifacts
- Can uncover bugs
- Unlike formal verification, simulation does not require knowledge of:
 - Temporal Logic, SAT modulo theories, Bounded Model Checking
 - Simulations are cheap and usually fast
 - Test-suites can be shared and built up across models
- Promising simulation-based approach: requirement falsification...



Requirement Falsification

Definition (Falsification):

Find parameters $p \in P$ and input $u \in U$ such that behaviors $\phi(\mathcal{M}, p, u)$ do NOT satisfy requirements φ (i.e., $\phi(\mathcal{M}, p, u) \neq \varphi$)



- Not verification, but systematic bug-finding
- No guarantees of completeness (except asymptotic/probabilistic)



Some key enablers

Robust satisfaction of φ by simulation trace $\phi(\mathcal{M}, p, u)$

- A function maps φ and $\phi(\mathcal{M}, p, u)$ to \mathbb{R}
- Positive number = $\phi(\mathcal{M}, p, u)$ satisfies ϕ
- Negative number = $\phi(\mathcal{M}, p, u)$ does not satisfy φ
- Moving towards zero = moving towards violation

Black-box Global optimizers

• Powerful heuristics to get close to global optimum



Falsification by optimization





 φ

Falsification supported by both **<u>S-TaLiRo</u>** and **<u>Breach</u>** tools

Requirement Falsification

• Work by others

- S-TaLiRo [Fainekos, Sankaranarayanan, et al.]
 - Metric Temporal Logic based requirements
 - Supports several stochastic optimizers
- Breach [Donzé, CAV 2010, NSV 2013]
 - Signal Temporal Logic based requirements
 - Supports Nelder-Mead
 - Can exploit sensitivity info



Requirement Falsification

- Other things we've done in the past
 - Multiple Shooting [Zutshi, Sankaranarayanan, et al., EMSOFT 2014, HSCC2016]
 - Multiple short simulations segments leading from initial conditions to unsafe states; adjust initial conditions to piece segments together
 - Stochastic Local Search for Falsification [with Deshmukh, et al., ATVA 2015]
 - Discrete optimization method used as search heuristic
 - Simulation-based testing for coverage [with Dreossi, et al., NASA Formal Methods 2015, extensions with Adimoolam, et al., CAV 2017]
 - · Selecting inputs to maximize coverage of infinite state-space
 - Simulation-based convergence/stability testing [with Sankaranarayanan et al., HSCC 2014, extensions with Balkan, et al., EMSOFT 2016]
 - Specifications in the form of Lyapunov-like function to test for convergence/stability



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Other Simulation-based Methods

- Other things we've done in the past
 - Simulation traces to learn contraction metrics [with Balkan, et al., ICC, 2015]
 - Simulations to learn Lyapunov-like function showing *convergence*; used to compute *flowpipes* that contain all system behaviors; used for verification (proving safety for infinite sets of behaviors)
 - Simulation-based verification [with Fan, et al., EMSOFT 2016]
 - Simulations used to compute *flowpipes* to prove safety
 - Simulation traces to assist mechanical theorem provers [with Arechiga, et al., EMSOFT 2015]
 - Simulations used to learn invariant sets; invariant sets used in theorem prover to show safety



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Image source: LinkedIn

REQUIREMENT ENGINEERING CHALLENGES



Requirement Engineering Challenges

- Outline
 - Overview of requirements engineering philosophy
 - Comparison of perspectives: software vs. CPS
 - Challenges
 - ST-Lib: collection of formal requirements for control engineering applications
 - Results and challenges applying ST-Lib





Requirements-Driven Approach

- Many of our efforts focus on providing a requirements-driven development approach
 - Requirements are developed and iterated on
 - Requirements used to develop control models and specify expected behaviors of models
 - Same requirements also used to define expected behaviors from calibration & test, as well as from the deployed system





Classic Verification Assumption







Takeaways:

- 1. Difficult/impossible to specify every aspect of CPS behaviors
- 2. Aspects of possible behaviors are discovered in simulation and testing phases



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The Reality for CPS





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Results from

CPS Requirement Challenges

- Requirements are evolving due to CPS-related issues
 - System hardware/software designs evolve concurrently
 - Not possible to create a plant model that captures all behaviors
 - Subtle interactions between states/signals are not known before integration test
- Definition of correct behaviors exist only in engineer's brain
 - Formal requirements are hard for engineers to develop
 - Existing requirements do not capture all of the desired behaviors
 - Model may capture appropriate/expected behavior but requirements do not



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ST-LIB SIGNAL TEMPLATE LIBRARY



• J. Kapinski, X. Jin, J. Deshmukh, A. Donze, T. Yamaguchi, H. Ito, T. Kaga, S. Kobuna, S. Seshia. "ST-Lib: A Library for Specifying and Classifying Model Behaviors". Society of Automotive Engineers Technical Paper (SAE), 2016.

What is ST-Lib?

- **IS** a library for specifying and classifying signal patterns of system behaviors
- **Isn't** a modeling language like Simulink for simulation



Why ST-Lib?

- **Can** formally specify intended design behaviors using a signal template
- Can automatically use simulation-based techniques to identify (near) worst-case behaviors of system



Introduction to STL

• Signal Temporal Logic (STL)

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- Specify timed behaviors of systems, containing:
 - Logic operators (\land , \neg , \lor , \rightarrow)
 - Temporal operators ("always", "eventually", and "until")
 - Atomic constraint formula $(f(x) \ge 0)$



ST-Lib

- ST-Lib uses STL to identify signal patterns of interest to design engineers, including:
 - Ringing
 - Spikes and glitches
 - Excessive overshoot or undershoot
 - Slow response time (settling, rising, or falling)
 - Undesirable timed relation behaviors
 - Steady state or tracking error



Example: Overshoot



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ST-LIB IN PRACTICE: LESSONS LEARNED AND CHALLENGES



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- Challenges applying ST-Lib in practice
 - Works well for simulation models and engineered input patterns
 - Does not work well with <u>real data</u> particularly real input patterns
 - Does not account for <u>subjective</u> nature of many evaluation practices





ST-Lib Example

- ST-Lib application example
 - Applied versions of the following ST-Lib templates to a fuel cell (FC) vehicle powertrain application
 - Overshoot
 - Settling time
 - Rise time
 - Steady-state error



ST-Lib Example

- ST-Lib application example
 - Example: used the following version of the overshoot requirement

OVERSHOOT := alw ((STEPUP and alw_[dt,sstime] not(STEP)) => alw_[dt,sstime] (OVERSHOOTLIMIT))

STEPUP := in[t+dt]-in[t] > StepThresh STEPDOWN := in[t]-in[t+dt] > StepThresh STEP := STEPUP or STEPDOWN

OVERSHOOTLIMIT := out[t] < 1.1*(in[t])

<u>Comment</u>:
No other step should be present when checking the overshoot



sstime: Time over which steady-state is assumed to be reached **dt**: Small constant, comparable to a sampling step size

- Overshoot requirement performance
 - Good (expected) requirement performance for control model, using engineered input patterns



Overshoot requirement performance

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- Bad (unexpected) requirement performance for real data



This behavior is appropriately identified as a fault





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- Problem: Overshoot error tolerance fixed
 - Engineer wants a.) relative error limit for large reference values and b.) absolute error limit for small reference values



- Problem: Many important behaviors are neither steps nor steadystate
 - These behaviors should fall under some other category of inputs, like an input *ramp*, with corresponding requirements



- Other challenges
 - Engineer had a notion of an ideal response (included a time-shifted, rate-limited version of the command signal)
 - Not easily captured in STL
 - Addressed by a priori defining a new signal
 - Using a fault localization (in time) tool
 - Very difficult for complex STL formulas



- ST-Lib shortcomings
 - Scaling constants relative to command magnitude
 - Step is not the only meaningful input
 - For real data, need to define a partition on the input and relate a corresponding behavioral constraint on output
 - Need to allow for more subjective classification of reference signal class (step, ramp, SS, others...)



- We created some STL-based solutions for all of the requirements, but...
 - The requirements represent approximations of what they actually want
 - The requirements do not capture faults with 100% accuracy (there are false positives/negatives)
 - The requirements are very complicated
 - Difficult to read/understand (this reduces the value of the requirement)
 - Are computationally expensive to monitor in Breach



- General formal requirements challenge
 - Subjective nature of behavior expectation difficult to capture with temporal logic (like STL)
 - We are capturing (poorly) right now using complex STL requirements
 - Need improved methods to capture designer intentions
 - Alternatives:
 - Write code that would monitor appropriate behaviors
 - » Downside: specification in the same language as the design (uses a program to specify correct behavior of a program)
 - Train a NN to classify (good/bad) behaviors like the engineer would
 - » **Downside**: Essentially provides a specification that is a black box
 - Other ideas?



Summary

- CPS is everywhere and is safety critical
 - Verification for CPS is hard!
 - New simulation-based analysis techniques
 - Simulation-based falsification methods can perform automated bug-finding
- Requirement engineering an ongoing challenges for CPS
 - ST-Lib intended to support V&V activities for CPS applications
 - Application results are promising but many challenges revealed
 - Need to think about improved methods to capture designer intentions



Other CPS Test & Verification Challenges

- Building appropriate models
 - Model creation is time consuming and error-prone
 - How to automate model construction
 - How to check model accuracy
- Verification techniques
 - Scaling model-checking/theorem proving techniques for CPS
 - Dealing with black-box models
- Advanced testing/evaluation techniques
 - Continue to develop new/better simulation-based falsification approaches
 - Need automated testing approaches for calibration
- Control synthesis
 - Can we create safe-by-construction control designs?
- Systems based on machine learning/AI
 - Lots of immediate applications: autonomous cars, advanced driver assist
 - Not clear how to test/certify





Thanks for your attention!

- Thanks to...
 - Toyota collaborators: Hisahiro Ito, Ken Butts, Jyotirmoy Deshmukh, Xiaoqing Jin
 - Academic collaborators: Alexandre Donzé, Tommaso Dreossi, Sanjit Seshia (UC Berkeley), Ayca Balkan, Paulo Tabuada (UCLA), Georgios Fainekos (Arizona State University), Nikos Arechiga, (Toyota Info Technology Center), Aditya Zutshi, Sriram Sankaranarayanan (CU Boulder)
- Special thanks to **Necmiye Ozay** for the invitation!
- Questions? Comments?





Please read about these and related issues in our article in the Dec. 2016 issue of IEEE Control System Magazine