The Science of Software and System Design

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Bridging the gap between DES and formal methods



Discrete Event Dynamic Systems

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Supervisory control and reactive synthesis: a comparative introduction

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Abstract

This paper presents an introduction to and a formal connection between synthesis problems for discrete event systems that have been considered, largely separately, in the two research communities of *supervisory control* in control engineering and *reactive synthesis* in computer

Cyber-physical systems: future



Courtesy https://vimeo.com/bsfilms
Thanks to Christos Cassandras for recommending this video

Cyber-physical systems: present



Autonomous car driving through red light

How do we typically design systems?

- The trial and error approach:
 - Build \rightarrow test \rightarrow fix \rightarrow repeat.
- Problems with this approach:
 - Un-scalable
 - Un-economic
 - Un-safe
- Yet common...

Are we the drivers supposed to debug the autopilot?

How to design **better systems**? How to **better design** systems?

"It was described as a beta release.

The system will learn over time and get better and that's exactly what it's doing. It will start to feel quite refined within a couple of months." – Elon Musk, Tesla CEO, Nov 2015

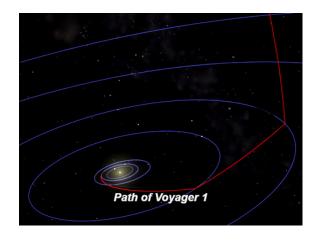
Tesla driver dies in first fatal crash while using autopilot mode June 2016

The autopilot sensors on the Model S failed to distinguish a white tractor-trailer crossing the highway against a bright sky



Is system design an art or a science?

- Science = knowledge that helps us make predictions
- We can make fairly good predictions about several systems we build (buildings, bridges, satellites, ...)
- What can we say about cyber-physical systems?
- What predictions can we make about software? (term used broadly)





Elements of the science of system design ("model-based design")

Simulink, UML, SysML, HDLs, SystemC, ... Describe the system that we want Modeling Simulation, verification, ... Be sure that this is what we want **Analysis Synthesis** Implement the system **Automatically** Correct-by-construction

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This talk: some recent work

The Refinement Calculus of Reactive Systems

 Synthesis of platform mappings with applications to security

- (Time permitting) Combining controller synthesis and learning
 - or Why model-based design is not the end of the story

The Refinement Calculus of Reactive Systems (RCRS)

Joint work with Viorel Preoteasa and Iulia Dragomir (Aalto)
Sponsors: Academy of Finland and NSF CPS Breakthrough

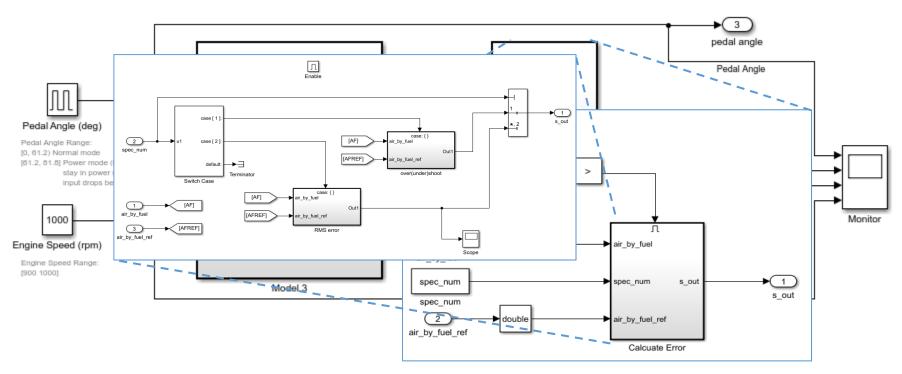


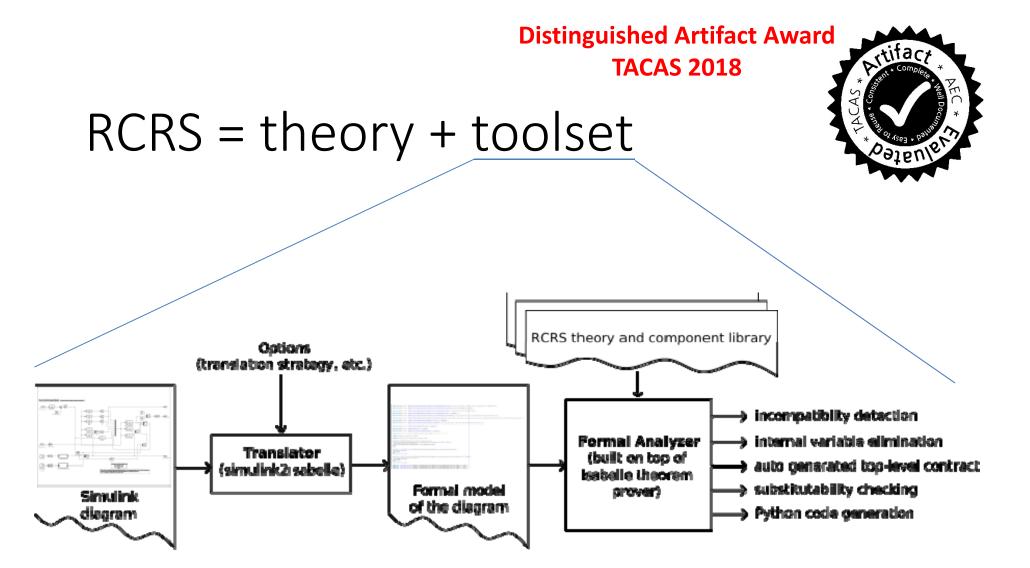




Motivation

• Compositional formal reasoning for CPS – Simulink:





Downloadable from http://rcrs.cs.aalto.fi/

RCRS theory: contract-based design

- Relational interfaces [EMSOFT'09, ACM TOPLAS'11]
 - Symbolic, synchronous version of interface automata [Alfaro, Henzinger]
 - Open, non-deterministic, non-input-complete systems (this is crucial for static analysis)
 - Semantic foundation: relations
 - Limited to safety properties
- Refinement calculus of reactive systems [EMSOFT'14]
 - Richer semantics: predicate and property transformers
 - Can handle both safety and liveness properties
 - Entirely formalized in Isabelle theorem prover 27k lines of Isabelle code

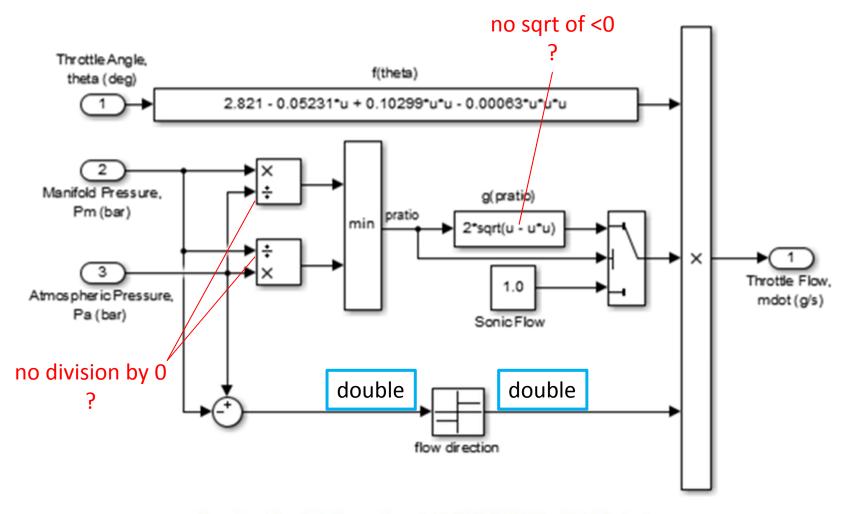
Some of the things RCRS can do

Design-by-contract

Incremental design with refinement

Compositional verification

Static ("compile-time") analysis

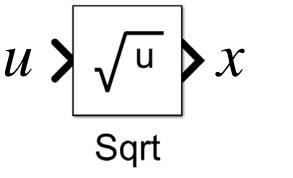


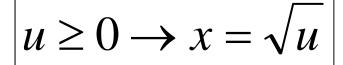
Simulink square root modeled with RCRS contracts



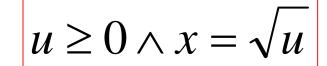
double -> double

Simulink type





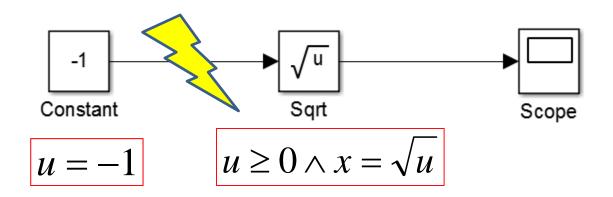
RCRS contract: input-receptive



RCRS contract:

non-input-receptive

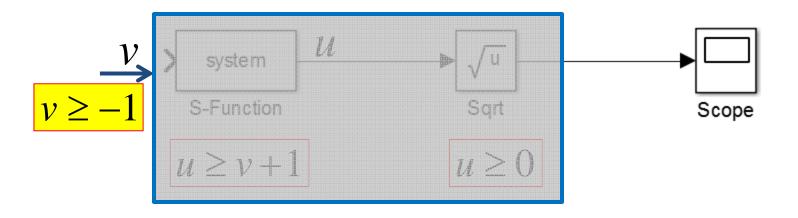
Catching incompatibilities statically



caught by taking the conjunction of the two formulas and checking satisfiability

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Inferring new contracts automatically



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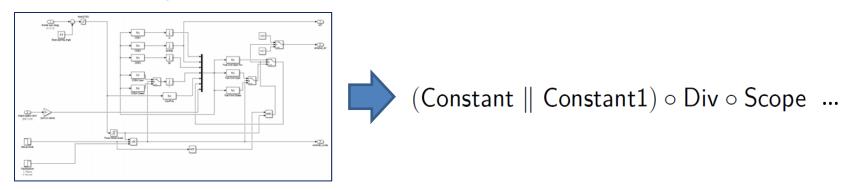
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Library of Simulink basic blocks in RCRS

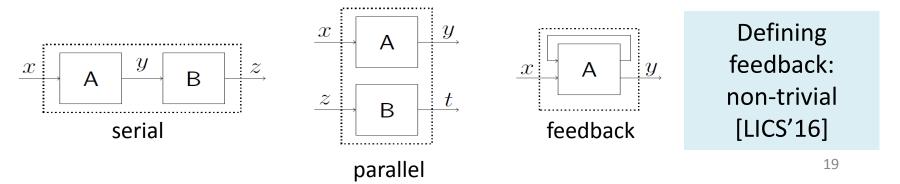
```
definition "Id = [: x \rightsquigarrow y . y = x :]"
definition "Add = [: (x, y) \rightsquigarrow z . z = x + y :]"
definition "Constant c = [: x::unit \rightsquigarrow y . y = c :]"
definition "UnitDelay = [: (x,s) \rightsquigarrow (y,s') . y = s \land s' = x :]"
definition "Sqrt = {. x . x \ge 0 .} o [: x \rightsquigarrow y . y = \sqrt{x} :]"
definition "NonDetSqrt = {. x . x \ge 0 .} o [: x \rightsquigarrow y . y \ge 0 :]"
definition "ReceptiveSqrt = [: x \rightsquigarrow y . x \ge 0 \implies y = \sqrt{x} :]"
definition "Integrator dt = [: (x,s) \rightsquigarrow (y,s'). y=s \land s'=s+x*dt :]"
```

Translation of (arbitrary) Simulink diagrams

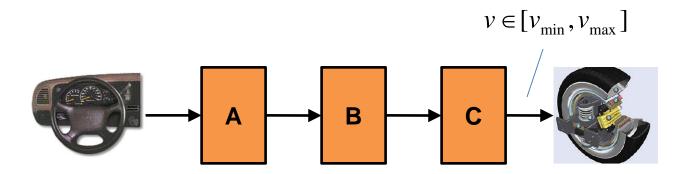
 Formal, modular, compositional translation: a non-trivial problem



- Algebra of block diagrams:
 - Only 3 composition primitives:



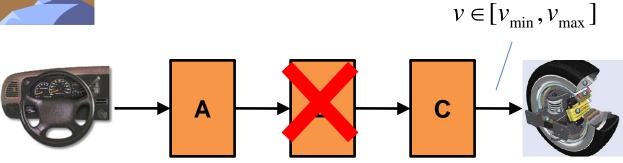
Suppose we have designed and verified this "steer-by-wire" system:

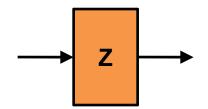


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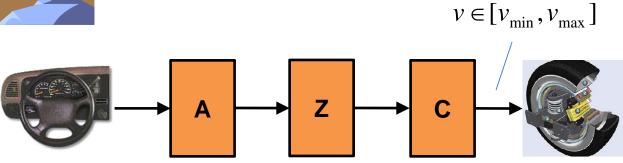
Suppose we want to replace B with Z:







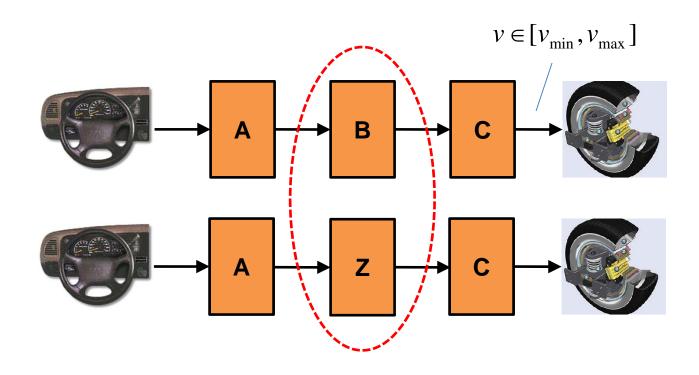
How to ensure properties are preserved (substitutability)?



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In RCRS it suffices to check that

Z ≤ B: Z refines B (local check)



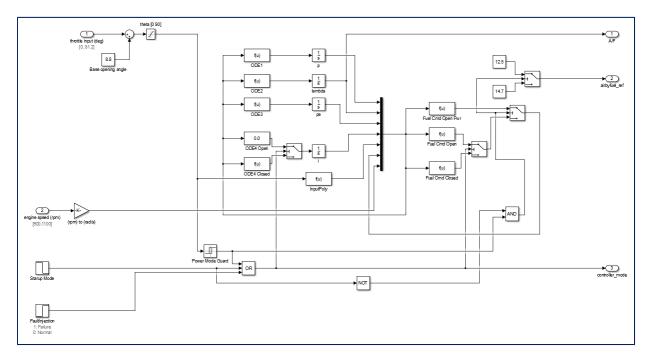
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Does it work for real-world systems?

- Case study: Fuel Control System automotive benchmark
- Made publicly available by Toyota on CPS-VO website
- Simulink model: 3-level hierarchy, 104 blocks
- Translator produces a 1660-line long RCRS theory (translation time negligible)
- Automatic static analysis / contract inference / simplification:

<1 minute

Sample subsystem of the FCS model



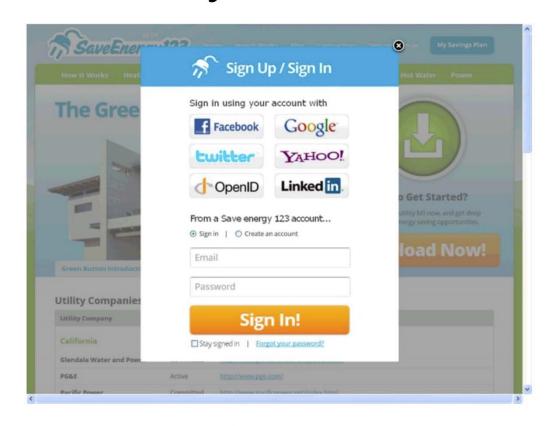
Synthesis of platform mappings with applications to security

Joint work with Eunsuk Kang (NSF ExCAPE project), and Stephane Lafortune (UMichigan)

Sponsors: NSF Expeditions ExCAPE

Motivation: security

Third-Party Authentication



OAuth: Widely adapted, support from major vendors Well-scrutinized & **formally checked**

Motivation: security



by Chris Brook May 2, 2014, 1:42 pm

UPDATE — A serious vulnerability in the OAuth and OpenID protocols could lead to complications for those who use the services to log in to websites like Facebook, Google, LinkedIn, Yahoo, and Microsoft among many others.

Study of OAuth providers [Sun & Beznosov, CCS12] Majority vulnerable (Google, Facebook,...)

The heart of the problem

Application Design Deployment

Platform

Designers think at high-level

Protocols, APIs, workflows, use cases, etc., Ignore irrelevant details

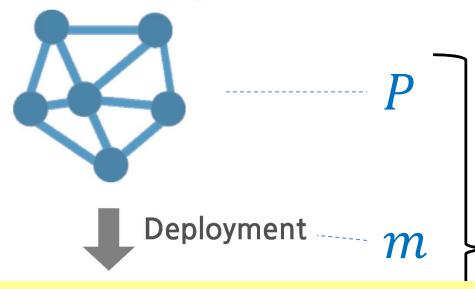
implementation

Attacks may exploit details absent at high-level

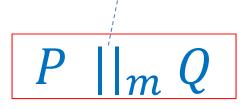
Unwanted features
Unknown environment
Hidden interface/entry points

Our approach: modular modeling with mappings

Application Design



mapping composition operator



Examples of decisions captured by mappings:

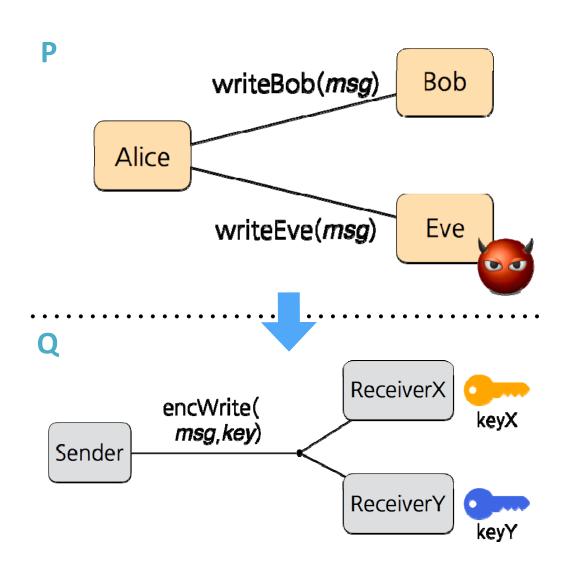
- should a certain protocol message be implemented as an HTTP request?
- with cookies to store secret values?
- with query parameters?

Possible applications beyond security.





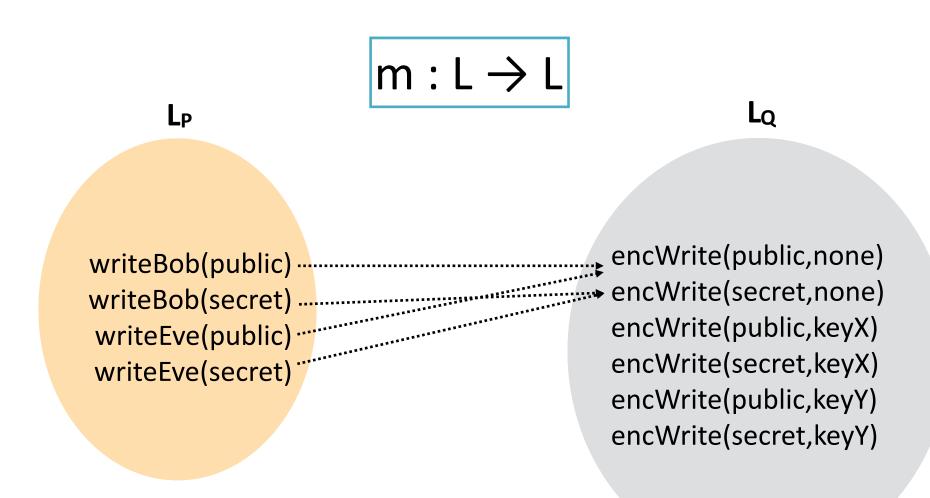
Example: abstract channel & public channel



S ≡ "Only Bob can learn Alice's secret"

Can we implement abstract design P on platform Q and preserve property \$?

Implementation decisions as mappings



Correct and incorrect mappings

m₁:

```
writeBob(secret) → encWrite(secret,none)
writeBob(public) → encWrite(public,none)
writeEve(secret) → encWrite(secret,none)
writeEve(public) → encWrite(public,none)
```

No messages encrypted Eve can read Alice's secret!

m₂:

```
writeBob(secret) → encWrite(secret, keyX)
writeBob(public) → encWrite(public, keyX)
writeEve(secret) → encWrite(secret, keyX)
writeEve(public) → encWrite(public, keyX)
```

Encrypt all messages (safe but inefficient)

m₃:

```
writeBob(secret) → encWrite(secret,keyX)
writeBob(public) → encWrite(public,none)
writeEve(secret) → encWrite(secret,keyX)
writeEve(public) → encWrite(public,none)
```

Public messages need no encryption

Verification and synthesis problems on mappings

• Verification: given application model P, platform model Q, mapping m, and some specification ϕ , check that the system $P \mid_{m} Q$ satisfies ϕ .

• Synthesis: given P, Q and ϕ , find mapping m, such that $P \mid_{m} Q$ satisfies ϕ .

Contributions

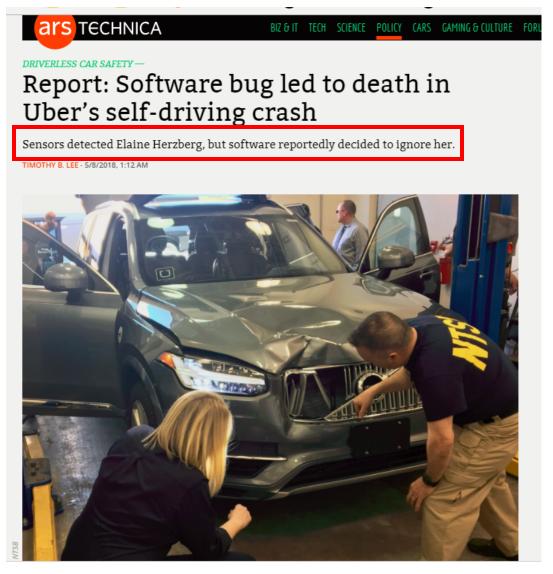
- Algorithm and tool for automated mapping synthesis:
 - Counter-example guided symbolic search over possible candidate mappings
- Real-world case studies: OAuth 2.0 and 1.0
 - Tool able to automatically synthesize correct mappings for both OAuth 2.0 and 1.0
 - Synthesized mappings describe mitigations to wellknown attacks (e.g., session swapping, covert redirect, session fixation)
 - Several 1000s LOC of application and platform models: OAuth, HTTP server, HTTP browser, ...

From Model-based to Data-driven and Model-based Design

Brave new world

Tempe, Arizona, March 18, 2018

- "Software designers face a basic tradeoff [...]. If the software is programmed to be too cautious, the ride will be slow and jerky [...]. Tuning the software in the opposite direction will produce a smooth ride most of the time—but at the risk that the software will occasionally ignore a real object. [...] that's what happened in Tempe in March and unfortunately the "real object" was a human being."
- "There's a reason Uber would tune its system to be less cautious about objects around the car, [...] It is trying to develop a self-driving car that is comfortable to ride in."



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New challenges and opportunities

Can AI benefit from system design, and how?

Can system design benefit from AI, and how?

Can Al benefit from system design?

- Yes.
- Al software is untestable.
- Formal verification of Al software is needed.



Driving to Safety

How Many Miles of Driving Would It Take to Demonstrate Autonomous Vehicle Reliability?

Nidhi Kalra, Susan M. Paddock

Key findings

- Autonomous vehicles would have to be driven hundreds of millions of miles and sometimes hundreds of billions of miles to demonstrate their reliability in terms of fatalities and injuries.
- Under even aggressive testing assumptions, existing fleets would take tens and sometimes hundreds of years

n the United States, roughly 32,000 people are killed and more than two million injured in crashes every year (Bureau of Transportation Statistics, 2015). U.S. motor vehicle crashes as a whole can pose economic and social costs of more than \$800 billion in a single year (Blincoe et al., 2015). And, more than 90 percent of crashes are caused by human errors (National Highway Traffic Safety Administration, 2015)—such as driving too fast and misjudging other drivers' behaviors, as well as alcohol impairment, distraction, and fatigue.

Can system design benefit from AI?

• Yes.

Data-driven and Model-based Design (DMD)

Data-driven and Model-based Design – motivation and goals

- Combine the best of both worlds:
 - Trial-and-error
 - Model-based design
- Leverage advances in AI (machine learning, data science, ...) to improve system design methods.
- Complement existing AI methods by developing new techniques developed specifically for system design.

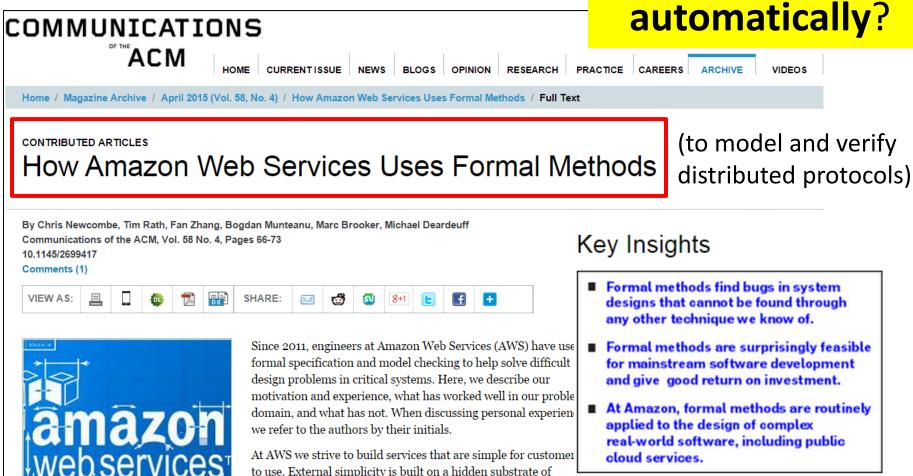
Example: combining controller synthesis and learning

Joint work with Rajeev Alur, Christos Stergiou et al (UPenn)
Sponsors: NSF Expeditions ExCAPE

Motivation: distributed protocols

Notoriously hard to get right

Can we synthesize such protocols automatically?



complex distributed greatenes. Quels complex internals are

Verification and synthesis in a nutshell

Verification:

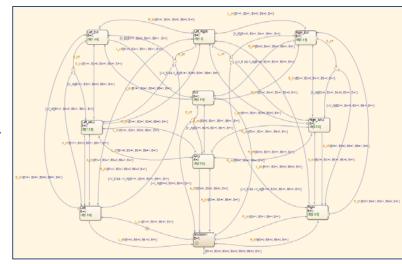
- 1. Design system "by hand": S
- 2. State system requirements: ϕ
- 3. Check: does S satisfy ϕ ?
- Synthesis (ideally):
 - 1. State system requirements: ϕ
 - 2. Generate automatically system S that satisfies ϕ by construction.

State of the art synthesis

From formal specs to discrete controllers:

```
#Assumptions
(gl_healthy & gr_healthy & al_healthy & ar_healthy)
[](gl_healthy | gr_healthy | al_healthy | ar_healthy)
[](!gl_healthy -> X(!gl_healthy) )
[](!gr healthy -> X(!gr healthy) )
[](!al_healthy -> X(!al_healthy) )
[](!ar_healthy -> X(!ar_healthy) )
#Guarantees
(!c1 & !c2 & !c3 & !c4 & !c5 & !c6 & !c7 & !c8 & !c9 & !c10 &
!c11 & !c12 & !c13)
[](X(c7) \& X(c8) \& X(c11) \& X(c12) \& X(c13))
[](!(c2 & c3))
[](!(c1 & c5 & (al_healthy | ar_healthy)))
[](!(c4 & c6 & (al_healthy | ar_healthy)))
[]((X(gl_healthy) & X(gr_healthy) ) -> X(!c2) & X(!c3) &
X(!c9) & X(!c10))
[]((X(!gl_healthy) & X(!gr_healthy) ) -> X(c9) & X(c10))
```





Specification (temporal logic formulas)

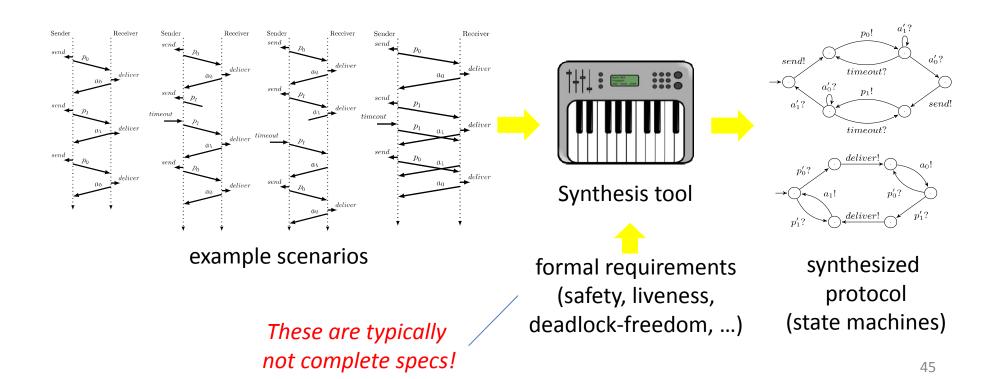
Controller (state machine)

• Limitations:

- Scalability (writing full specs & synthesizing from them)
- Not applicable to distributed protocols (undecidable)

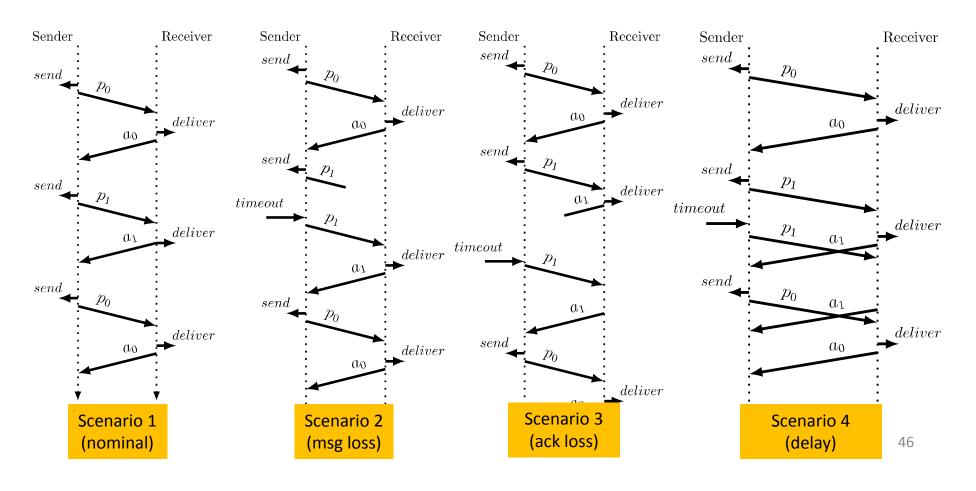
Synthesis of Distributed Protocols from Scenarios and Requirements

• Idea: combine requirements + example scenarios



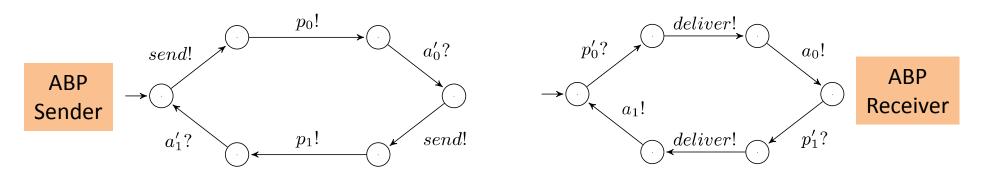
Scenarios: message sequence charts

- Describe what the protocol must do in some cases
- Intuitive language ⇒ good for the designer
- Only a few scenarios required (1-10)

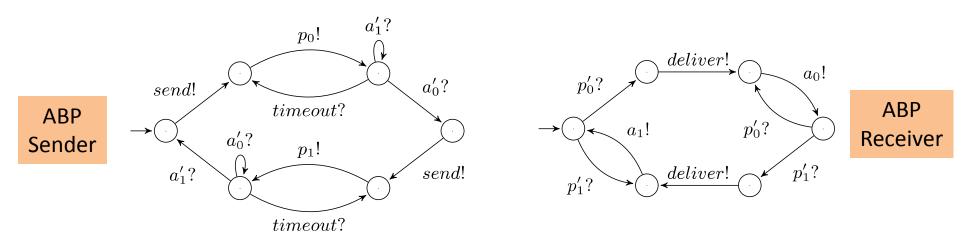


Synthesis becomes a completion problem

Incomplete automata learned from first scenario:

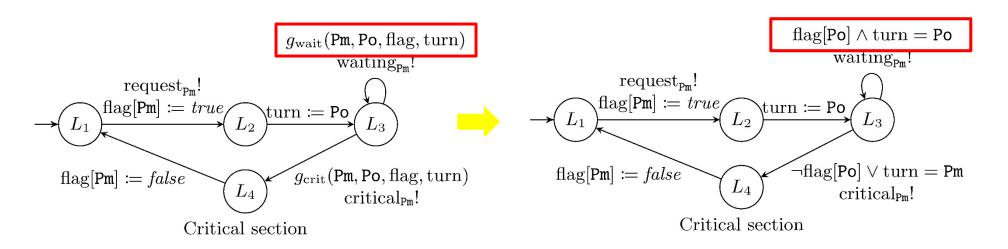


Automatically completed automata:



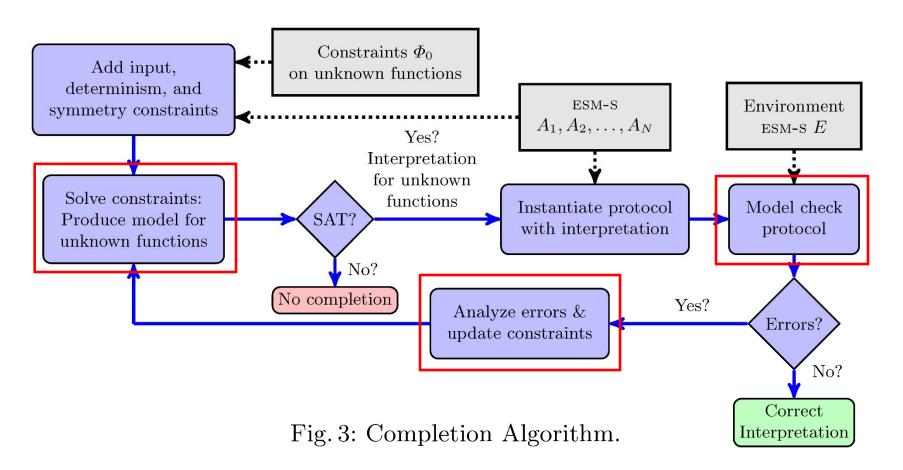
Results

- Able to synthesize the distributed Alternating Bit Protocol (ABP) and other simple finite-state protocols (cache coherence, consensus, ...) fully automatically [HVC'14, ACM SIGACT'17].
- Towards industrial-level protocols described as extended state machines [CAV'15].



Algorithmic technique: counter-example guided completion of (extended) state machines

 Completion of incomplete machines: find missing transitions, guards, assignments, etc.



Combining synthesis with learning

- Synthesis: given specification ϕ , find system S, such that $S \models \phi$
- Learning: given set of examples E, find system S, such that S is consistent with E and "generalizes well" ...
- Synthesis from spec + examples: given set of examples E and specification ϕ , find system S, such that S is consistent with E and $S \models \phi$
 - Key advantage: ϕ guides the generalization!

CONCLUSIONS

The science of system design

- Theory and tools that help us make better predictions about the systems we build.
- Formal modeling, verification, synthesis, ...
 - A.k.a. "formal methods".
- Broad spectrum of interesting research problems (theory and practice).
 - Increasingly mature for education.
- Increasingly popular in the industry.
- New opportunities: data, examples, learning!

Thank you

Questions?