Automatically Robustifying Verified Hybrid Systems in KeYmaera X

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Robustness

A system is **robust** if it operates correctly despite:

- Disturbances in actuation
- Uncertainty in sensing
- Deviation from typical dynamics
- Adversarial agents
- . .

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Expressible by systematically modifying a hybrid system

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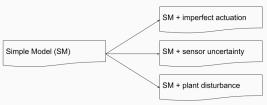
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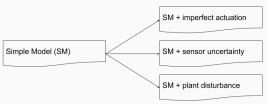
Expressible by systematically modifying a hybrid system

Can we automatically robustify hybrid systems?

Typical verification approach: begin with a **simplified model**, then incrementally add **complexity**.



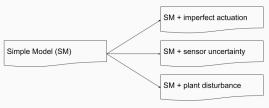
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Advantages:

- Initial verification task exposes essential aspects of the safety argument.
- Successive verification tasks are tractable.

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Goal: Automatic Incremental Robustification

Specifying Hybrid Systems

Definition (Hybrid Programs)

```
Assign x:=\theta
Sequence \alpha; \beta
Test ?\varphi
Iteration \alpha^*
Choice \alpha \cup \beta
```

ODEs $\{x'_1 = \theta_1, \dots, x'_n = \theta_n \& H\}$

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Differential Dynamic Logic (d \mathcal{L}) formulas describe reachability properties of hybrid programs using modalities: $[\alpha]\varphi$ and $\langle\alpha\rangle\varphi$.

Specifying Hybrid Systems



```
[{
 \{?(x \ge \frac{(AT + v)^2}{2B} + obs); a := A \cup a := -B \}; 
 c := 0; \{x' = v, v' = a, c' = 1 \land v \ge 0 \land c \le T \} 
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- Parametric controller design
- Non-determinism

$$A > 0 \land B > 0 \land T > 0 \land v \ge 0 \land \frac{v^{2}}{2B} + obs \le x \le obs$$

$$\to [\{\{(x \ge \frac{(AT + v)^{2}}{2B} + obs); a := A \cup a := -B \};$$

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- Parametric controller design
- Non-determinism
- Symbolic constraints on parameters

Verifying a Simple Hybrid System in KeYmaera X

KeYmaera X is a **trustworthy** and **scriptable** hybrid systems theorem prover.

- Trustworthy: All prover automation passes through a small soundness-critical core (< 2 KLOC).
- Scriptable: KeYmaera X provides a DSL for writing proof search programs.

Example: Adding Actuation Error

$$A > 0 \land B > 0 \land T > 0 \land v \ge 0 \land$$

 $\frac{v^2}{2B} + obs \le x \le obs \rightarrow$
 $\{\{(x \ge \frac{((A)T + v)^2}{2(B)} + obs); a := A \cup a := -B\};$
 $c := 0; \{x' = v, v' = a, c' = 1 \land v \ge 0 \land c \le T\}$
 $\{x' \le obs\}$

Example: Adding Actuation Error

$$A > 0 \land B > 0 \land T > 0 \land v \ge 0 \land 0 < \epsilon < A \land \epsilon < B \land \frac{v^2}{2B \pm \epsilon} + obs \le x \le obs \rightarrow$$
[{
$$\{?(x \ge \frac{((A \pm \epsilon)T + v)^2}{2(B \pm \epsilon)} + obs\}; a := A \pm \epsilon \cup a := -B \pm \epsilon\};$$

$$c := 0; \{x' = v, v' = a, c' = 1 \land v \ge 0 \land c \le T\}$$

$$\}^*]x \le obs$$

Example: Adding Actuation Error

$$A > 0 \land B > 0 \land T > 0 \land v \ge 0 \land 0 < \epsilon < A \land \epsilon < B \land \frac{v^2}{2B-\epsilon} + obs \le x \le obs \rightarrow \{\{(x \ge \frac{((A+\epsilon)T+v)^2}{2(B-\epsilon)} + obs); a := A+\epsilon \cup a := -B-\epsilon\}; c := 0; \{x' = v, v' = a, c' = 1 \land v \ge 0 \land c \le T\} \}^*]x \le obs$$

Co-Transformation of Models and Tactics

Simple Model

```
ImplyR(1) & loop(p(x,v,a,A,B),
1) <(
    QE, QE,
    splitCases(1) <(
        chase(1) & ODE & QE
        chase(1) & ODE & QE</pre>
```

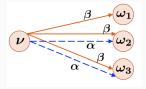
Simple Model + Uncertainty

```
\begin{split} & \mathsf{ImplyR}(1) \ \& \\ & \mathsf{loop}(\mathsf{p}(\mathsf{x},\mathsf{v},\mathsf{a},\!\mathsf{A}\!+\!\epsilon,\!\mathsf{B}\!-\!\epsilon),\ 1) < (\\ & \mathsf{QE},\ \mathsf{QE},\\ & \mathsf{splitCases}(1) < (\\ & \mathsf{chase}(1) \ \& \ \mathsf{ODE} \ \& \ \mathsf{QE}\\ & \mathsf{chase}(1) \ \& \ \mathsf{ODE} \ \& \ \mathsf{QE}\\ & \mathsf{))} \end{split}
```

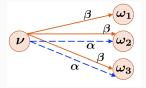
Incremental Robustification via Model/Proof Co-Transformation

- √ Tractable initial verification
- √ Verification of robustified models re-use ideas from initial safety proof
 - ? Compositional robustification
- √ Re-verification is expensive (manual effort)
- × Re-verification is expensive (computationally)

System α refines system β ($\alpha \leq \beta$) if every state reachable by α is also reachable by β .

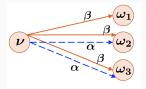


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 Many robustifications are refinements (after changing environment and controller).

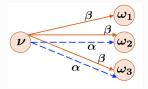
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- Refinement makes *direct* use the initial safety property:

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• < has a well-understood algebraic structure.

Conclusions and Further Thoughts

Automatic incremental robustification automates common changes to CPS models

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Further Thoughts:

- It would be nice to have automatic robustification procedures for high-fidelity models of common sensors and actuators.
- Notions of robustness are describable in differential game logic (dGL); automation story is unclear.

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Thanks: KeYmaera X developers (Stefan Mistch, Andrè Platzer, Brandon Bohrer, Jan-David Quesel)

Advertisement: KeYmaera X Tutorial at FM this year!