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# Delta-Complete Reachability Analysis

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Apr 22, 2011

# Hybrid Systems

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$$\mathcal{H} = \langle X, Q, Init, Flow, Jump, Inv \rangle$$

- $X \subseteq \mathbb{R}^k$ : state space
- $Q$ : a finite set of modes
- $Init \subseteq Q \times X$ : initial configurations
- $Flow : \subseteq Q \times X \rightarrow TX$ : continuous flows
- $Jump : \subseteq Q \times X \rightarrow 2^{Q \times X}$ : discrete jumps
- $Inv \subseteq Q \times X$ : invariants in each mode

Given  $Unsafe \subseteq \mathbb{R}^k \times Q$ ,  $[[\mathcal{H}]] \cap Unsafe = \emptyset$ ?

# Example

## Example (Transmission Controller)

- $X = \mathbb{R}^3$  ( $v$ : Speed,  $Th$ : Throttle,  $Fr$ : Friction)
- $Q = \{q_1, q_2, q_3\}$  (Gears)
- $Init = (q_1, Th = 0.2 \wedge v = 0)$
- $Inv_{q_1} : 0 \leq v \leq 30, Inv_{q_2} : 25 \leq v \leq 50, Inv_{q_3} : 45 \leq v \leq 70.$
- $Flow_{q_i} : \frac{dv}{dt} = c_i(a_i Th - b_i Fr) \wedge \frac{dFr}{dt} = e_i v^2.$
- $Jump_{q_1, q_2} : (v \geq 20 \wedge Th > 0.6 \wedge v' = v \wedge Th' = Th), \text{ etc.}$

Is  $(q_2, Th = 0.1 \wedge v < 30)$  reachable?

# Hybrid System Verification is Hard.

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- Although there are successful examples, most of the practical systems can not be handled.
- Main Approaches:
  - Geometric Methods
    - 1 Over-estimate  $\llbracket H \rrbracket$  up to some time bound  $t$ .
    - 2 Check if  $\llbracket H \rrbracket^{<t} \cap Unsafe = \emptyset$ .
  - Proof-theoretic Methods
    - 1 Show that  $\Phi(\mathcal{H}) \vdash \neg Unsafe$  is derivable syntactically in a sound axiomatic system.

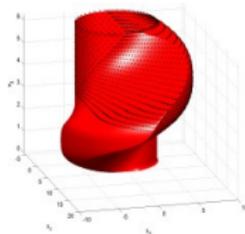
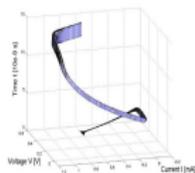
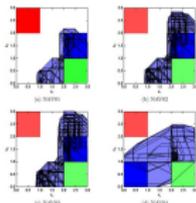
# Geometric Methods

## Pros:

- Computations can be made visible.
- Very helpful for the general understanding of behavior.

## Cons:

- High complexity; error control is hard.
- Hard to handle complex dynamics or high dimensions.
- Hard to handle logical operations.



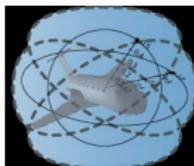
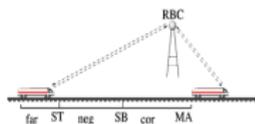
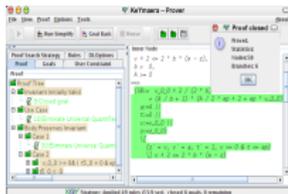
# Proof-theoretic Approaches

## Pros:

- Highly complex systems/properties.
- Reliable answers.
- No bounds on variables.

## Cons:

- Not for debugging.
- Finding invariants needs much human insight.
- Underlying decision procedures can be hard to scale.



# Stepping Back

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What made traditional model checking scale?

- Encode verification problems into logic formulas.
  - View  $\langle X, \rightarrow \rangle$  as a logical structure.
  - Encode properties of interest as a temporal/propositional formula  $\varphi$ .
- Check *satisfiability* of formulas using highly efficient solvers.
  - $\langle X, \rightarrow \rangle \models \varphi$ ?
  - Use BDD/SAT/SMT solvers to find a model of  $\varphi$ .

Comparison: debugging information, flexible; not visible, bounded

# Model-theoretic Methods (Discrete Systems)

Let  $\mathcal{M}$  denote the transition system  $\langle X, \rightarrow \rangle$ .

## ■ Bounded Reachability

$$\mathcal{M} \models \exists \vec{x}_0, \dots, \vec{x}_n (Init(\vec{x}_0) \wedge \bigwedge_{i=0}^{n-1} Trans(\vec{x}_i, \vec{x}_{i+1}) \wedge Target(\vec{x}_n))?$$

## ■ Reachable Set Computation

$$\llbracket \exists \vec{x}_0, \dots, \vec{x}_{n-1} (Init(\vec{x}_0) \wedge \bigwedge_{i=0}^{n-1} Trans(\vec{x}_i, \vec{x}_{i+1})) \rrbracket^{\mathcal{M}} = ?$$

## ■ Synthesis Problems

$$\llbracket \forall \vec{x}_0, \dots, \vec{x}_n (Init(\vec{x}_0) \wedge \bigwedge_{i=0}^{n-1} Control(\vec{x}_i, \vec{x}_{i+1}, \vec{u}_i) \wedge Target(\vec{x}_n)) \rrbracket^{\mathcal{M}} = ?$$

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# Model-theoretic Methods (Continuous Systems)

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Logical encoding is not limited to discrete systems.

■ **Continuous Dynamics:**  $\frac{d\vec{x}(t)}{dt} = \vec{f}(\vec{x}(t), t)$

■ The solution curve:

$$\alpha : \mathbb{R} \rightarrow X, \alpha(t) = \alpha(0) + \int_0^t \vec{f}(\alpha(s), s) ds.$$

■ Define the predicate

$$\llbracket Flow_f(\vec{x}_0, \vec{x}, t) \rrbracket^{\mathcal{M}} = \{(\vec{x}_0, \vec{x}, t) : \alpha(0) = \vec{x}_0, \alpha(t) = \vec{x}\}$$

■ **Reachability**

$$\mathcal{M} \models \exists \vec{x}_0, t, \vec{x} (Init(\vec{x}_0) \wedge Flow_f(\vec{x}, \vec{x}_0, t) \wedge Target(\vec{x})) ?$$

# Model-theoretic Methods (Hybrid Systems)

Combine the discrete and continuous components<sup>1</sup>:

■  $Reach_{q_0 \rightarrow q_0}^0(\vec{x}) :$

$$\exists t_0 \exists \vec{x}_0 (Inv_{q_0}(\vec{x}_0) \wedge Inv_{q_0}(\vec{x}) \wedge Flow_{q_0}(\vec{x}, \vec{x}_0, t_0))$$

■  $Reach_{q_0 \rightarrow q}^{n+1}(\vec{x}) :$

$$\begin{aligned} & \exists t_{n+1} \exists \vec{x}_{n+1} \exists \vec{x}'_{n+1} \\ & \bigvee_{q' \in Q} [Reach_{q_0 \rightarrow q'}^n(\vec{x}_{n+1}) \wedge Jump_{q' \rightarrow q}(\vec{x}_{n+1}, \vec{x}'_{n+1}) \\ & \quad \wedge Flow_q(\vec{x}, \vec{x}'_{n+1}, t_{n+1}) \wedge Inv_q(\vec{x}') \wedge Inv_q(\vec{x}'_{n+1})] \end{aligned}$$

$$\mathcal{H} \models Reach_{q_0 \rightarrow q}^{n+1}(\vec{x}) \wedge Unsafe(\vec{x})?$$

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<sup>1</sup> Assumption: In each location, the flow stays within the invariant before any jump. 🔍 🔔 🔍 🔔

# Decision Procedures over Reals

Sadly, in general those first-order formulas over  $\mathbb{R}$  can never be decided.

- The arithmetic theory ( $\times/+$ ) is decidable but highly complex (double-exponential, PSPACE).
  - Available solvers: Usually hard to scale to more than 10 variables.
- Handling nontrivial systems will involve (in the *Flow* predicate)  $\exp, \sin / \cos, ODEs, \dots$ 
  - Wildly undecidable.

# Allowing Errors

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On the other hand, large systems of real equalities/inequalities/ODEs are routinely solved **numerically**.

- They are perfect for simulation, but always regarded inappropriate for verification.
  - (Platzer and Clarke, HSCC 2008)
- Is there a way of using them still?

# Allowing Errors

Decide

$$\exists \vec{x}. f(\vec{x}) = 0 \wedge g(\vec{x}) = 0.$$

- **Symbolically:** We need to consider the global algebraic properties of  $f$  and  $g$ .
- **Numerically:** We use iterations that only involve local evaluations of  $f$  and  $g$  (and their derivatives).
- With error bound  $\delta$ , we'd “numerically” decide:

$$\exists \vec{x}. |f(\vec{x})| < \delta \wedge |g(\vec{x})| < \delta.$$

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# Robust Formulas

Consider any formula

$$\varphi := \exists^{I\vec{x}} \vec{x}. \bigvee \left( \bigwedge_i f_i(\vec{x}) = 0 \wedge \bigwedge_j g_j(\vec{x}) \neq 0 \right)$$

Define its  $\delta$ -perturbed form

$$\varphi^\delta := \exists^{I\vec{x}} \vec{x}. \bigvee \left( \bigwedge_i f_i(\vec{x}) < \delta \wedge \bigwedge_j g_j(\vec{x}) \geq \delta \right)$$

We say  $\varphi$  is  $\delta$ -robust iff

$$\varphi \leftrightarrow \varphi^\delta.$$

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# Robust Formulas (Decidability)

Robust formulas have very nice computational properties.

## Definition

$\mathbb{R}_{\mathcal{F}} = \langle \mathbb{R}, \mathcal{F}, < \rangle$  where  $\mathcal{F}$  is the set of all real-computable functions. (Type-II computability; exp, sin, ODEs...)

Let  $\varphi$  be a robust and bounded sentence (arbitrary quantification):

## Theorem

$\mathbb{R}_{\mathcal{F}} \models \varphi$  is decidable.

The proof simulates cylindrical decomposition.

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# Robust Formulas (Complexity)

In particular, if  $\varphi$  is existentially quantified:

## Theorem

*If  $\mathcal{F}|\varphi$  is real-computable in complexity class  $\mathcal{C}$ , then deciding  $\varphi$  is in  $NP^{\mathcal{C}}$ .*

This means:

## Corollary

*Deciding robust bounded existential sentences*

- 1 in  $\mathcal{L}_{+, \times, \exp, \sin}$  is NP-complete.*
- 2 in  $\mathcal{L}_{\text{LipschitzODE}}$  is PSPACE-complete.*

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# Not Just in Theory

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We are developing the practical SMT solver **dReal**.

- DPLL(T) + Interval Constraint Propagation.
  - SAT solver handles Boolean skeleton, ICP handles systems of equations (scalable to  $10^3$  variables) .
- Currently solvable signature:  $+/\times$ , **exp**, **sin**
  - (Gao et al. FMCAD2010)
- In progress: (numerically stable) nonlinear ODEs

# Interval Constraint Propagation

- Interval Arithmetic + Constraint Solving

## Example

Solve  $\{x = y, x^2 = y\}$  for  $x \in [1, 4]$  and  $y \in [1, 5]$ :

$$I^x : [1, 4] \rightarrow [1, \sqrt{5}] \rightarrow [1, \sqrt[4]{5}] \rightarrow [1, \sqrt[8]{5}] \rightarrow [1, \sqrt[16]{5}] \rightarrow \dots \rightarrow [1, 1]$$

$$I^y : [1, 5] \rightarrow [1, \sqrt{5}] \rightarrow [1, \sqrt[4]{5}] \rightarrow [1, \sqrt[8]{5}] \rightarrow [1, \sqrt[16]{5}] \rightarrow \dots \rightarrow [1, 1]$$

- ICP routinely handles thousands of variables and highly nonlinear constraints.

# Correctness Guarantee (Formula)

For any existential formula  $\varphi$  (robust or nonrobust), with a tunable error bound  $\delta$ , we know:

- 1 Solver says “**unsat**”  $\Rightarrow \varphi$  is  $\delta$ -robustly unsatisfiable.
  - Unsatisfiable under any perturbation up to  $\delta$ .
- 2 Solver says “**sat**”  $\Rightarrow \varphi$  may be unsatisfiable, but  $\varphi^\delta$  is satisfiable.
  - It means we do know that a syntactically-perturbed version of  $\varphi(\vec{x})$  is satisfiable.

This is what we call  $\delta$ -completeness.

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# Robust Hybrid Systems

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Let  $\mathcal{H} = \langle X, Q, Init, Flow, Jump, Inv \rangle$ .

Similarly, we can define  $\delta$ -robust hybrid systems:

- $\mathcal{H}^\delta = \langle X, Q, Init^\delta, Flow^\delta, Jump^\delta, Inv^\delta \rangle$

- $\mathcal{H}$  is  $\delta$ -robust if

$$\mathcal{H} \sim_{\sigma.bisim} \mathcal{H}^\delta$$

# Delta-Complete Bounded Model Checking

When model checking  $\mathcal{H}$ :

$$\varphi : Reach_{\mathcal{H}}^{\leq n} \text{ is unsat} \Leftrightarrow \mathcal{H} \text{ is safe up to } n$$

- 1  $\varphi$  is “unsat”  $\Rightarrow \mathcal{H}$  is  $\delta$ -robustly safe.
  - $\mathcal{H}^{\vec{c}}$  is safe under any  $\delta$ -perturbation  $\vec{c}$ .
- 2  $\varphi$  is “sat”  $\Rightarrow \exists \delta$ -perturbation  $\vec{c}$ ,  $\mathcal{H}^{\vec{c}}$  is unsafe.
  - The solver returns a solution that shows bug.

This is even better than precise solvers!

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# Delta-Complete Bounded Model Checking

## Pros:

- Highly scalable numerical algorithms and SAT solvers
- Possible to scale to complex dynamics and large dimensions
- No accumulation of numerical errors
- Strong robustness check
- Counterexamples

## Cons:

- Bounded variables (can be very loose)
- Bounded unwinding depth
- Computations are not visible
- Debugging, not verifying (yet!)

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# Conclusion

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- Standard model checking techniques (from HW/SW) can be used in realistic continuous/hybrid systems, as long as we have the solver.
- For any solver to scale in this domain, numerical methods have to be exploited.
- Surprisingly, numerical methods will give us stronger results.
- We are developing `dReal` and `dReach`.

“Errors are good (if they work for the verification side).”