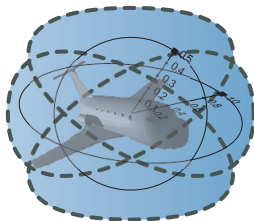


The Logical Path to Autonomous Cyber-Physical Systems

André Platzer

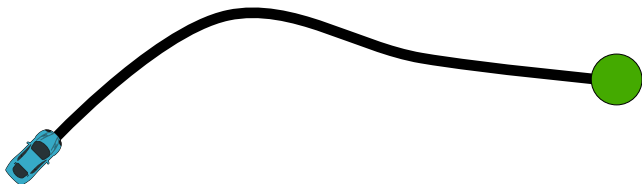
Carnegie Mellon University



- 1 Autonomous Cyber-Physical Systems
- 2 Foundation: Differential Dynamic Logic
- 3 ModelPlex: Model Safety Transfer
- 4 VeriPhy: Executable Proof Transfer
- 5 Safe Learning in CPSs
- 6 Applications
 - Airborne Collision Avoidance System
 - Ground Robot Navigation
- 7 Summary

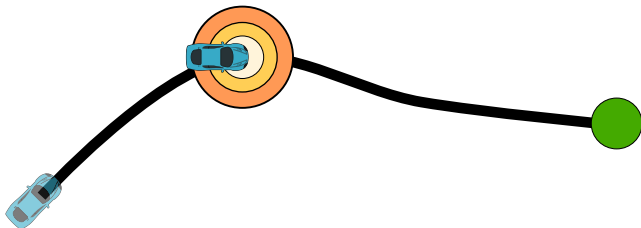


- 1 Autonomous Cyber-Physical Systems
- 2 Foundation: Differential Dynamic Logic
- 3 ModelPlex: Model Safety Transfer
- 4 VeriPhy: Executable Proof Transfer
- 5 Safe Learning in CPSs
- 6 Applications
 - Airborne Collision Avoidance System
 - Ground Robot Navigation
- 7 Summary



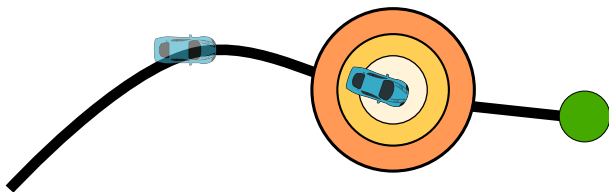
Cyber-Physical Systems

CPSs combine cyber capabilities with physical capabilities to solve problems that neither part could solve alone.



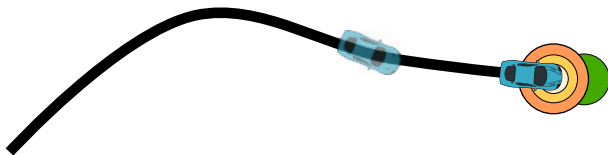
Cyber-Physical Systems

CPSs combine cyber capabilities with physical capabilities to solve problems that neither part could solve alone.



Cyber-Physical Systems

CPSs combine cyber capabilities with physical capabilities to solve problems that neither part could solve alone.



Cyber-Physical Systems

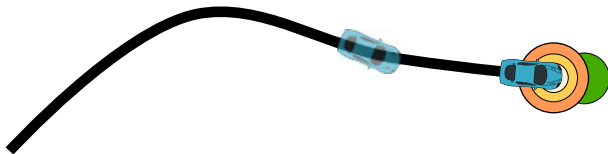
CPSs combine cyber capabilities with physical capabilities to solve problems that neither part could solve alone.

CPS Analysis

- Simple control
- ODE model
- Strong predictions
- Nondet decisions

AI Learning

- Flexible responses
- “No” model*
- Hard to predict
- Optimal decision ($t \rightarrow \infty$)



Cyber-Physical Systems

CPSs combine cyber capabilities with physical capabilities to solve problems that neither part could solve alone.

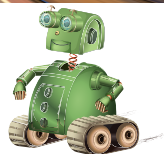
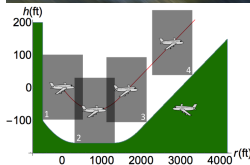
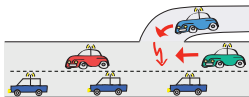
Autonomous CPSs Are on the Rise Everywhere

Prospects: Safety & Efficiency & Autonomy

Autonomous cars

Autonomous pilots

Robots near humans



Objective

Best of both worlds: safety from CPS + flexibility from AI



- 1 Autonomous Cyber-Physical Systems
- 2 Foundation: Differential Dynamic Logic**
- 3 ModelPlex: Model Safety Transfer
- 4 VeriPhy: Executable Proof Transfer
- 5 Safe Learning in CPSs
- 6 Applications
 - Airborne Collision Avoidance System
 - Ground Robot Navigation
- 7 Summary

Autonomous CPS



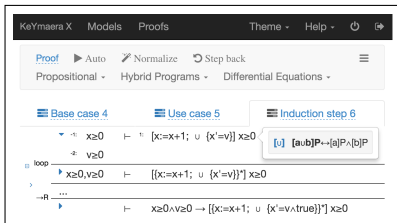
Monitor transfers safety

ModelPlex proof synthesizes

Compliance Monitor

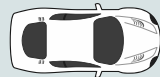


KeYmaera X



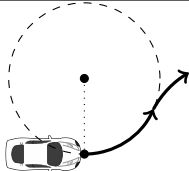
generates proofs

Proof and invariant search



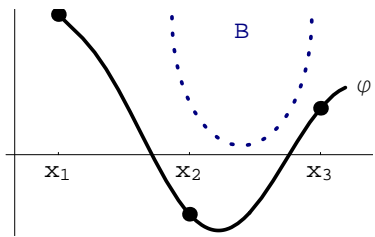
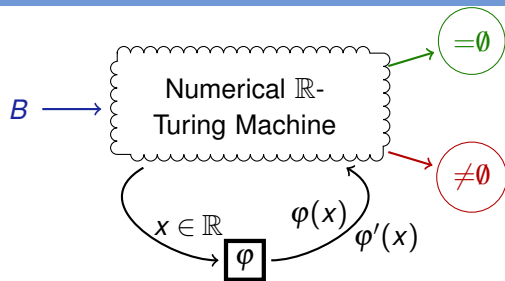
Model Safety

Model

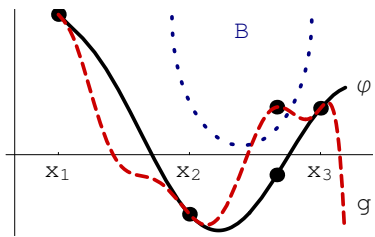
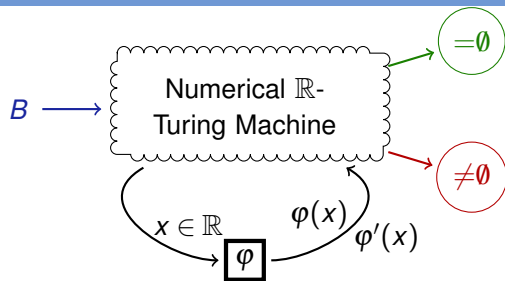


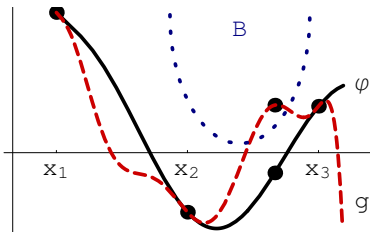
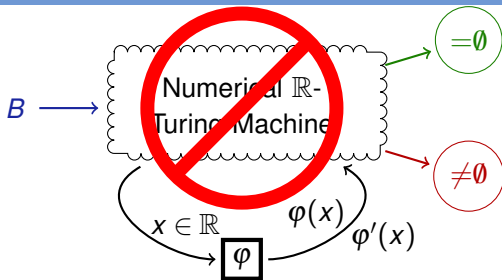
actions: $\{acc, brake\}$
motion: $x'' = a$

A Model Requirements



A Model Requirements

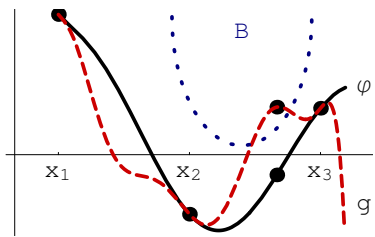
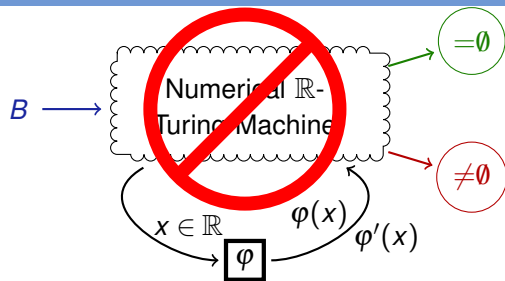




Proposition (Continuous image computation undecidable)

$\varphi(D) \cap B \stackrel{?}{=} \emptyset$ is undecidable by evaluating $\varphi(x)$ for

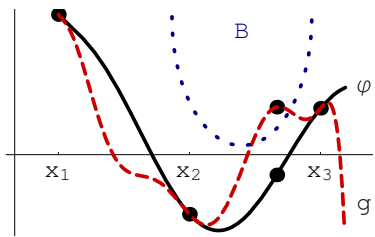
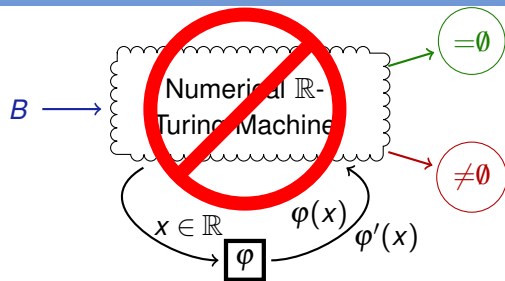
- arbitrarily effective flow $\varphi \in C^k(D \subseteq \mathbb{R}^n, \mathbb{R}^m)$ with effective D, B
- even if tolerating error $\varepsilon > 0$ in decisions



Proposition (Continuous image computation undecidable)

$\varphi(D) \cap B \stackrel{?}{=} \emptyset$ is undecidable by evaluating $\varphi(x)$ for

- arbitrarily effective flow $\varphi \in C^k(D \subseteq \mathbb{R}^n, \mathbb{R}^m)$ with effective D, B
- even if tolerating error $\varepsilon > 0$ in decisions
- even φ smooth polynomial function with \mathbb{Q} -coefficients
- even in Blum-Shub-Smale “real Turing machines”



Proposition (Continuous image computation undecidable)

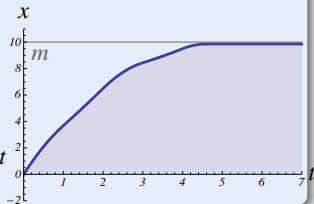
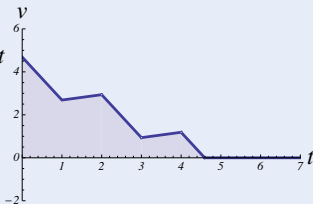
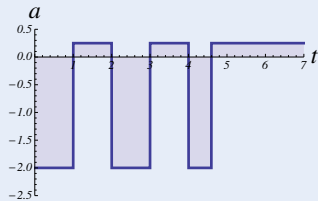
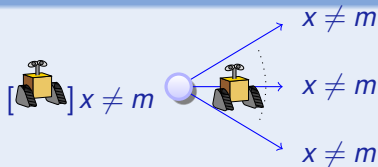
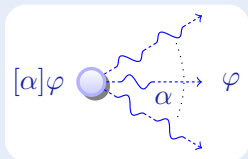
$\varphi(D) \cap B \stackrel{?}{=} \emptyset$ is undecidable by evaluating $\varphi(x)$ for

- arbitrarily effective flow $\varphi \in C^k(D \subseteq \mathbb{R}^n, \mathbb{R}^m)$ with effective D, B
- even if tolerating error $\varepsilon > 0$ in decisions
- even φ smooth polynomial function with \mathbb{Q} -coefficients
- even in Blum-Shub-Smale “real Turing machines”

The promise of “no model” is a myth

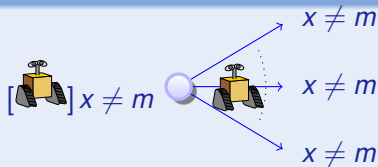
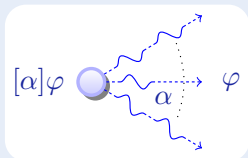
Concept (Differential Dynamic Logic)

(JAR'08, LICS'12)



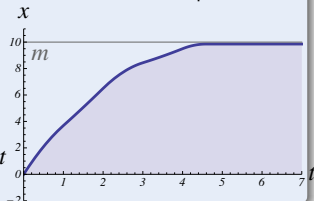
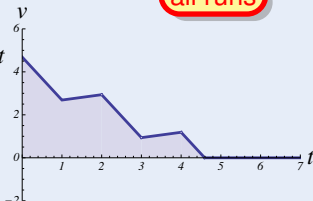
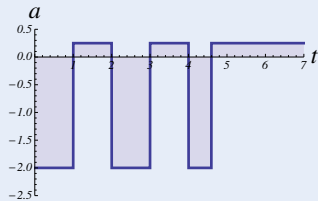
Concept (Differential Dynamic Logic)

(JAR'08, LICS'12)



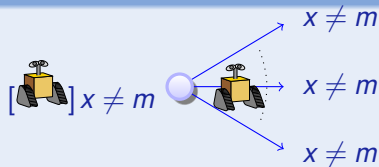
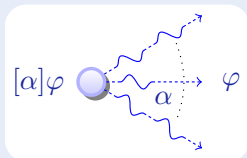
$$\left[\left(\text{if}(\text{SB}(x, m)) \quad a := -b \right); x' = v, v' = a \right]^* x \neq m \quad \text{post}$$

all runs



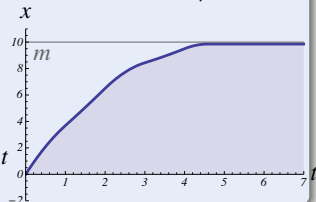
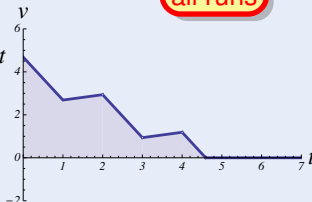
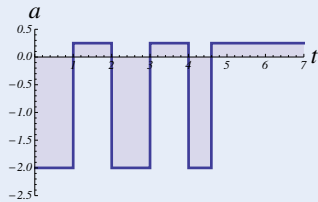
Concept (Differential Dynamic Logic)

(JAR'08, LICS'12)



$$\underbrace{x \neq m \wedge b > 0}_{\text{init}} \rightarrow \left[\left(\text{if}(\text{SB}(x, m)) \quad a := -b \right); x' = v, v' = a \right]^* \underbrace{x \neq m}_{\text{post}}$$

all runs



$$[:=] [x := e]P(x) \leftrightarrow P(e)$$

equations of truth

$$[?] [?Q]P \leftrightarrow (Q \rightarrow P)$$

$$['] [x' = f(x)]P \leftrightarrow \forall t \geq 0 [x := y(t)]P \quad (y'(t) = f(y))$$

$$[\cup] [\alpha \cup \beta]P \leftrightarrow [\alpha]P \wedge [\beta]P$$

$$[;] [\alpha; \beta]P \leftrightarrow [\alpha][\beta]P$$

$$[*] [\alpha^*]P \leftrightarrow P \wedge [\alpha][\alpha^*]P$$

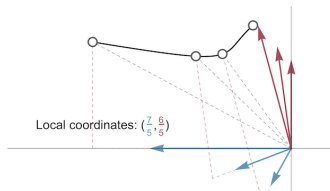
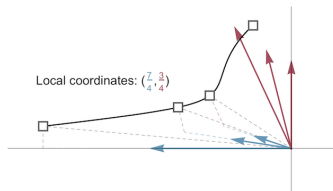
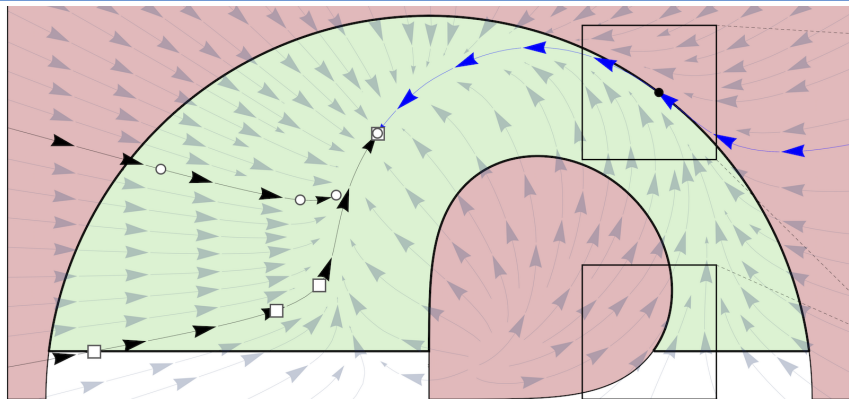
$$\mathbb{K} [\alpha](P \rightarrow Q) \rightarrow ([\alpha]P \rightarrow [\alpha]Q)$$

laws of logic of
laws of physics

$$\mathbb{I} [\alpha^*]P \leftrightarrow P \wedge [\alpha^*](P \rightarrow [\alpha]P)$$

$$\mathbb{C} [\alpha^*]\forall v > 0 (P(v) \rightarrow \langle \alpha \rangle P(v-1)) \rightarrow \forall v (P(v) \rightarrow \langle \alpha^* \rangle \exists v \leq 0 P(v))$$

Completeness for Differential Equation Invariants



Theorem (Algebraic Completeness)

(LICS'18, JACM'20)

dL calculus is a sound & complete axiomatization of algebraic invariants of polynomial differential equations. They are decidable by DI, DC, DG in dL.

Theorem (Semialgebraic Completeness)

(LICS'18, JACM'20)

dL calculus with RI is a sound & complete axiomatization of semialgebraic invariants of differential equations. They are decidable in dL.

Theorem (Algebraic Completeness) (LICS'18,JACM'20)

dL calculus is a sound & complete axiomatization of algebraic invariants of polynomial differential equations. They are decidable

$$\text{DRI } [x' = f(x) \& Q]e = 0 \leftrightarrow (Q \rightarrow e'^* = 0) \quad (Q \text{ open})$$

Theorem (Semialgebraic Completeness) (LICS'18,JACM'20)

dL calculus with RI is a sound & complete axiomatization of semialgebraic invariants of differential equations. They are decidable

$$\text{SAI } \forall x (P \rightarrow [x' = f(x)]P) \leftrightarrow \forall x (P \rightarrow P'^*) \wedge \forall x (\neg P \rightarrow (\neg P)^{*-})$$

Definable e'^* is short for *all/significant* Lie derivative w.r.t. ODE

Definable e'^{*-} is w.r.t. backwards ODE $x' = -f(x)$. Also for P .

$$e'^* = 0 \equiv e=0 \wedge (e')'^* = 0 \quad (P \wedge Q)^{ '* } \equiv P'^* \wedge Q'^*$$

$$e'^* \geq 0 \equiv e \geq 0 \wedge (e=0 \rightarrow (e')'^* \geq 0) \quad (P \vee Q)^{ '* } \equiv P'^* \vee Q'^*$$



- 1 Autonomous Cyber-Physical Systems
- 2 Foundation: Differential Dynamic Logic
- 3 ModelPlex: Model Safety Transfer**
- 4 VeriPhy: Executable Proof Transfer
- 5 Safe Learning in CPSs
- 6 Applications
 - Airborne Collision Avoidance System
 - Ground Robot Navigation
- 7 Summary

Autonomous CPS



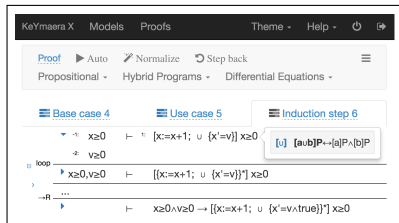
Monitor transfers safety

ModelPlex proof synthesizes

Compliance Monitor

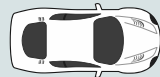


KeYmaera X



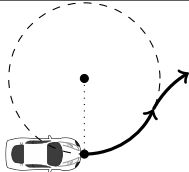
generates proofs

Proof and invariant search



Model Safety

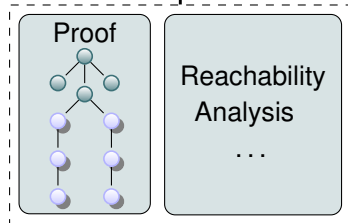
Model



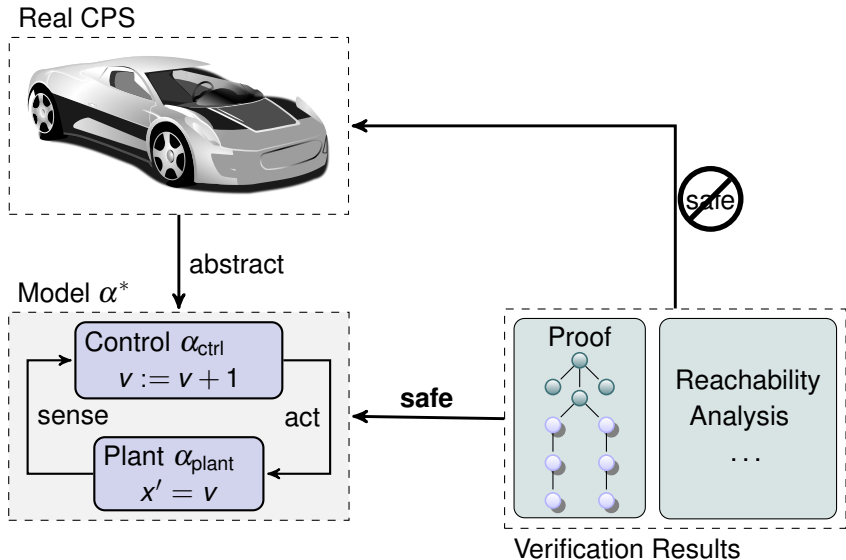
Real CPS



safe



Verification Results



Real CPS

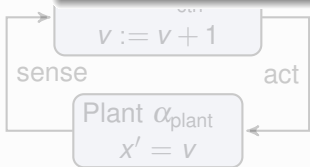


Challenge

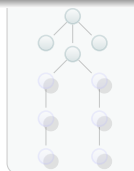
Verification results about models
only apply if CPS fits to the model

↷ Verifiably correct runtime model validation

Model



safe

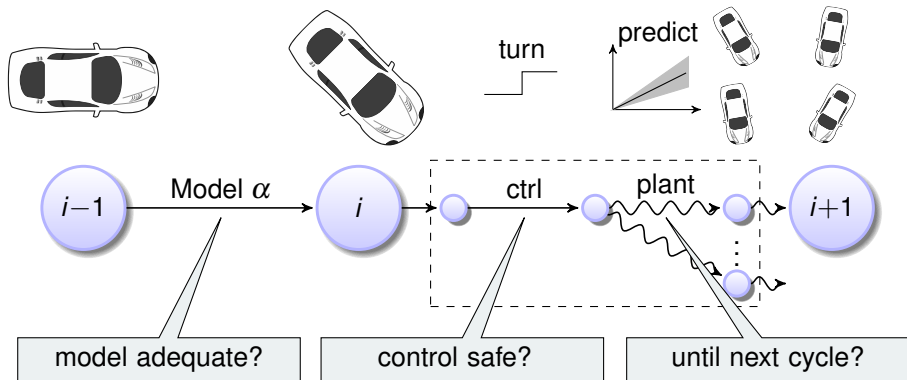


Reachability
Analysis

...

Verification Results

ModelPlex **ensures that verification results** about models **apply to CPS implementations**



ModelPlex **ensures that verification results** about models
apply to CPS implementations

Insights

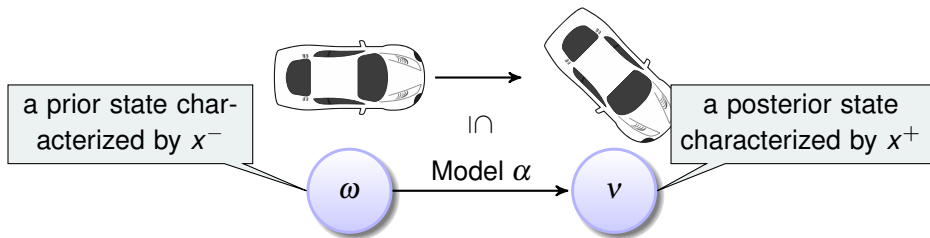
- Verification results about models transfer to the CPS when validating model compliance.
- Compliance with model is characterizable in logic dL.
- Compliance formula transformed by dL proof to monitor.
- Correct-by-construction provably correct model validation at runtime.

model adequate?

control safe?

until next cycle?

When are two states linked through a run of model α ?

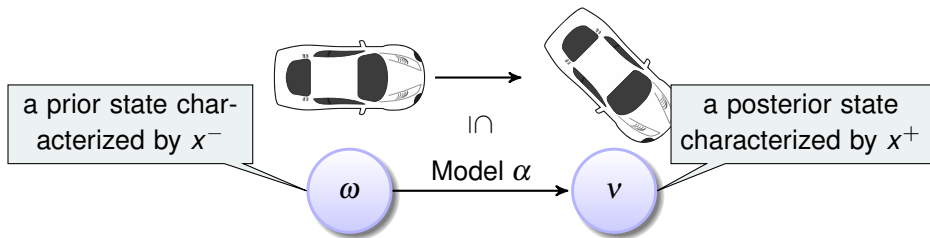


Semantical:

$$(\omega, \nu) \in \llbracket \alpha \rrbracket$$

reachability relation of α

When are two states linked through a run of model α ?



Offline

Semantical:

$$(\omega, \nu) \in \llbracket \alpha \rrbracket$$

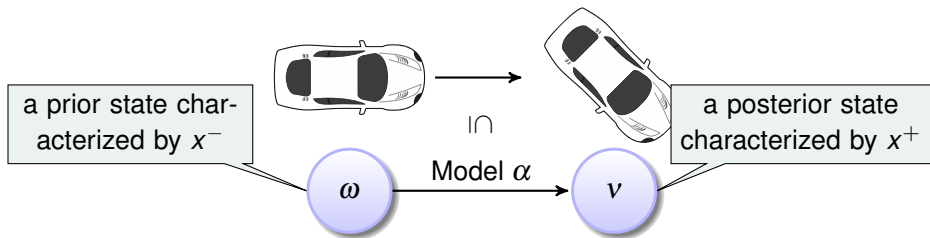
\Updownarrow Lemma

Logical dL:

$$(\omega, \nu) \models \langle \alpha \rangle (x = x^+)$$

exists a run of α to a state where $x = x^+$

When are two states linked through a run of model α ?



Offline

Semantical: $(\omega, \nu) \in \llbracket \alpha \rrbracket$

\Updownarrow Lemma

Logical dL: $(\omega, \nu) \models \langle \alpha \rangle (x = x^+)$

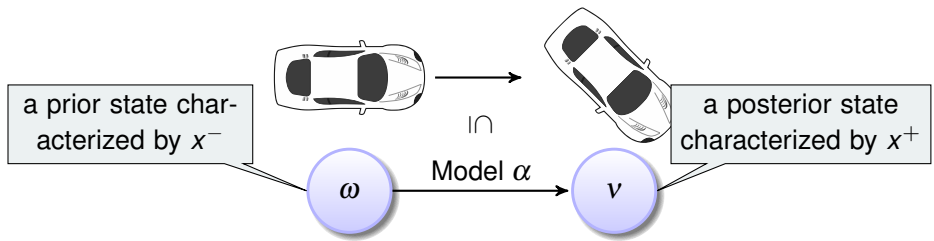
\Updownarrow dL proof

Arithmetical: $(\omega, \nu) \models F(x^-, x^+)$

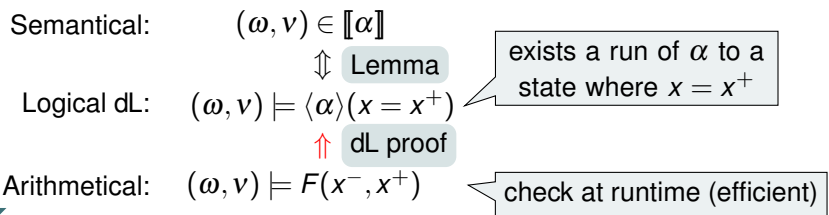
exists a run of α to a state where $x = x^+$

check at runtime (efficient)

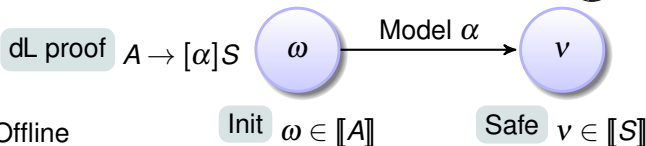
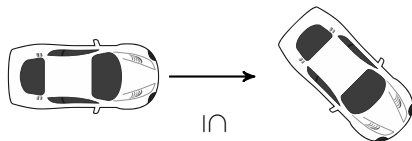
When are two states linked through a run of model α ?



Offline



Logic reduces CPS safety to runtime monitor with offline proof



Offline

Semantical: $(\omega, v) \in \llbracket \alpha \rrbracket$

\Updownarrow Lemma

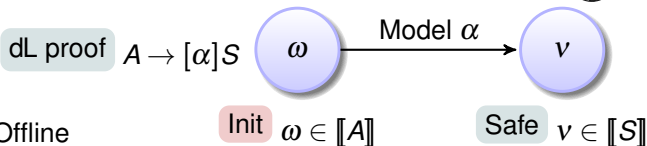
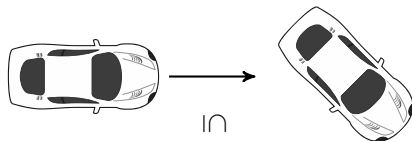
Logical dL: $(\omega, v) \models \langle \alpha \rangle (x = x^+)$

\Uparrow dL proof

Arithmetical: $(\omega, v) \models F(x^-, x^+)$

check at runtime (efficient)

Logic reduces CPS safety to **runtime** monitor with offline proof



Offline

Semantical: $(\omega, v) \in \llbracket \alpha \rrbracket$

\Downarrow Lemma

Logical dL: $(\omega, v) \models \langle \alpha \rangle (x = x^+)$

\Uparrow dL proof

Arithmetical: $(\omega, v) \models F(x^-, x^+)$

check at runtime (efficient)



- 1 Autonomous Cyber-Physical Systems
- 2 Foundation: Differential Dynamic Logic
- 3 ModelPlex: Model Safety Transfer
- 4 VeriPhy: Executable Proof Transfer**
- 5 Safe Learning in CPSs
- 6 Applications
 - Airborne Collision Avoidance System
 - Ground Robot Navigation
- 7 Summary

Autonomous CPS



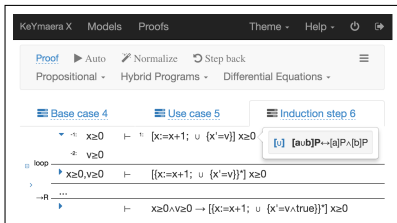
Monitor transfers safety

ModelPlex proof synthesizes

Compliance Monitor

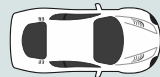


KeYmaera X



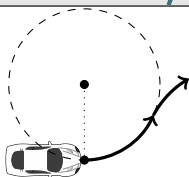
generates proofs

Proof and invariant search

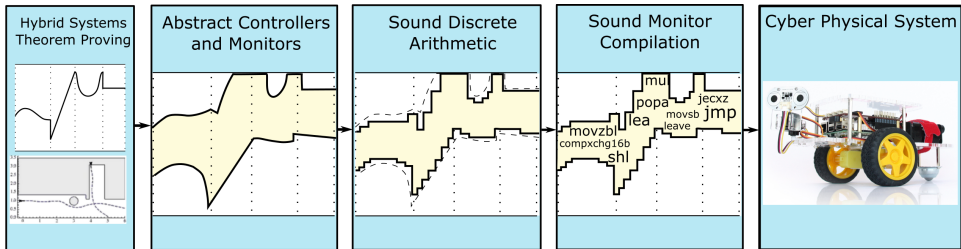


Model Safety

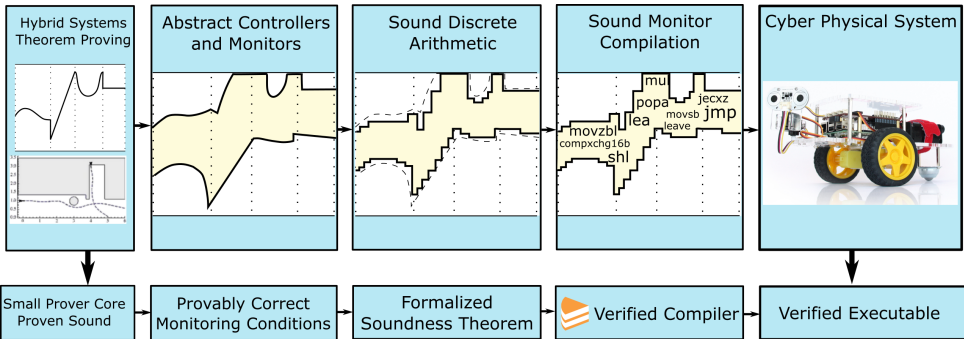
Model



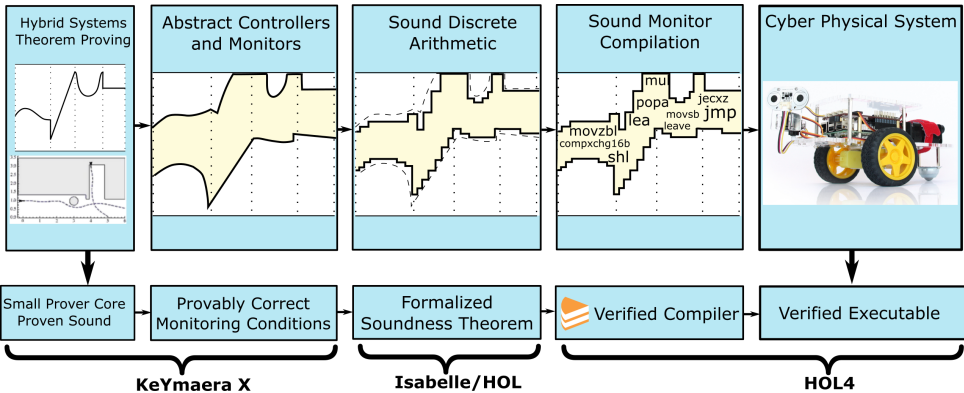
actions: $\{acc, brake\}$
motion: $x'' = a$



VeriPhy: Automatic, Verified EXEs from Controllers



VeriPhy: Automatic, Verified EXEs from Controllers



Your
Model

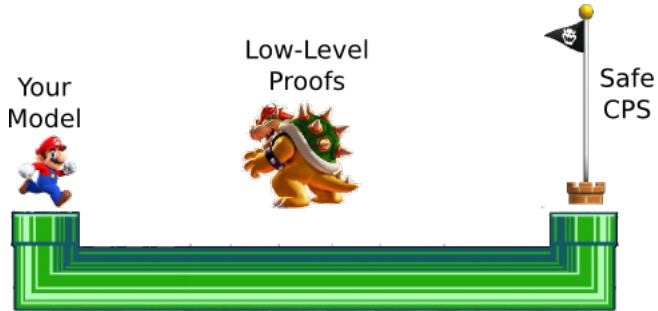


Low-Level
Proofs



Safe
CPS





VeriPhy Pipeline (VeriPhy.org)



- 1 Autonomous Cyber-Physical Systems
- 2 Foundation: Differential Dynamic Logic
- 3 ModelPlex: Model Safety Transfer
- 4 VeriPhy: Executable Proof Transfer
- 5 Safe Learning in CPSs**
- 6 Applications
 - Airborne Collision Avoidance System
 - Ground Robot Navigation
- 7 Summary

Autonomous CPS



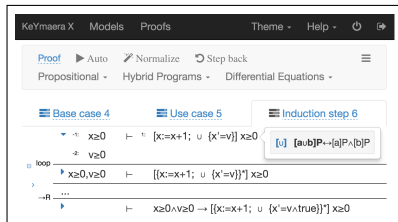
Monitor transfers safety

ModelPlex proof synthesizes

Compliance Monitor

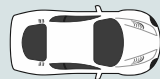


KeYmaera X



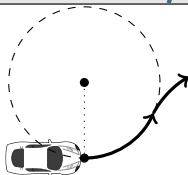
generates proofs

Proof and invariant search

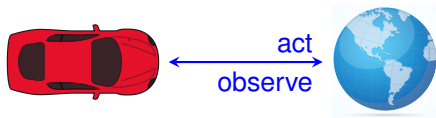


Model Safety

Model



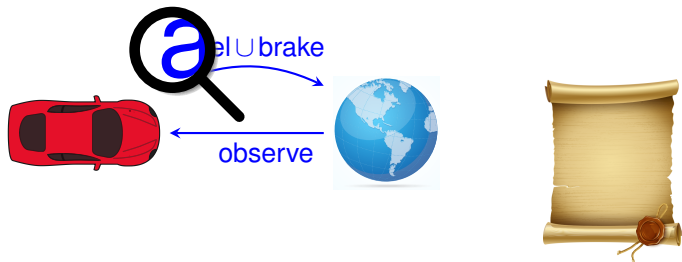
actions: $\{acc, brake\}$
motion: $x'' = a$



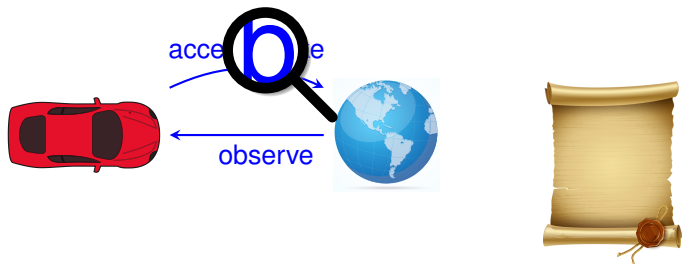
Reinforcement Learning learns from experience of trying actions



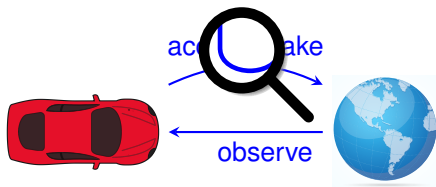
RL chooses an action, observes outcome, reinforces in policy if successful



ModelPlex monitor inspects each decision, vetoes if unsafe

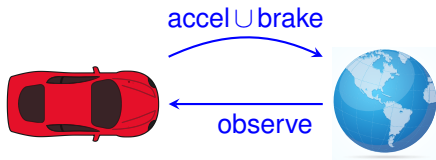


ModelPlex monitor gives early feedback about possible future problems.
No need to wait till disaster strikes and propagate back.



dL benefits from RL optimization.

RL benefits from dL safety signal.



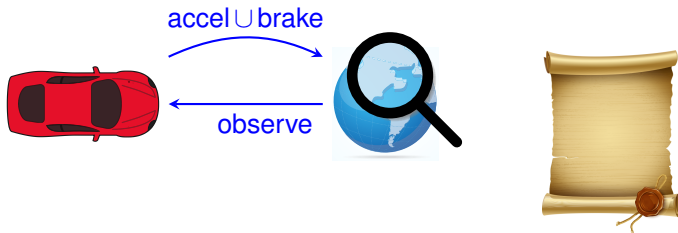
Theorem

Safe policy if ODE accurate

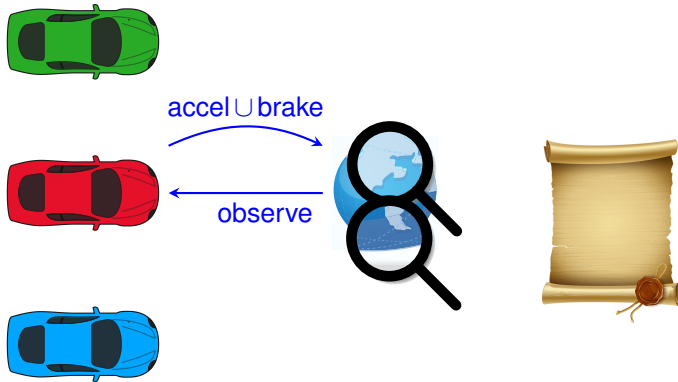
Experiment

Graceful recovery outside ODE \Leftarrow quantitative ModelPlex

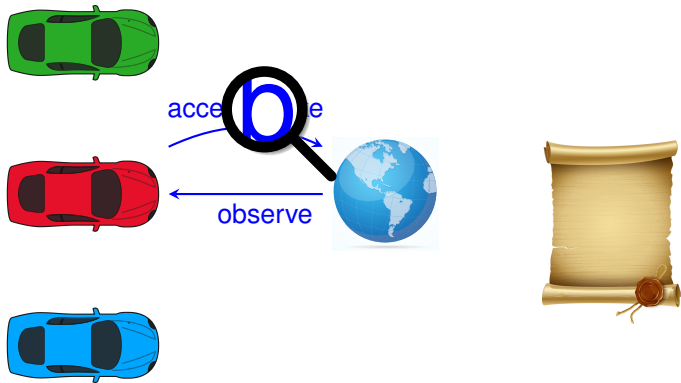
Detect modeled versus unmodeled state space \Leftarrow ModelPlex



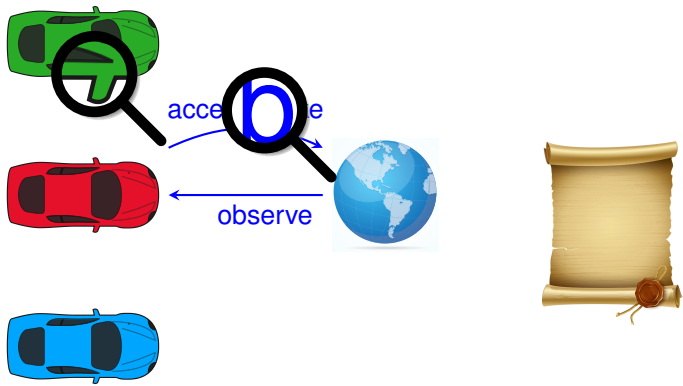
What's safe when off model?



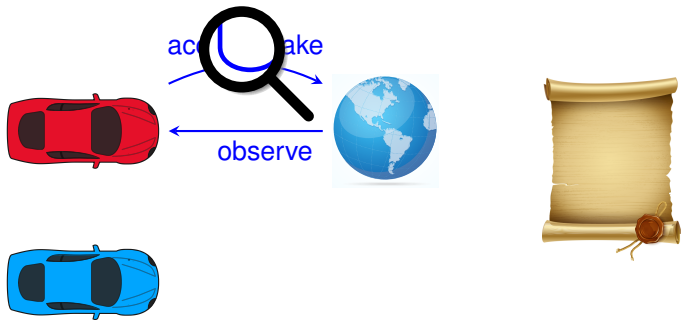
What's safe with multiple possible models?



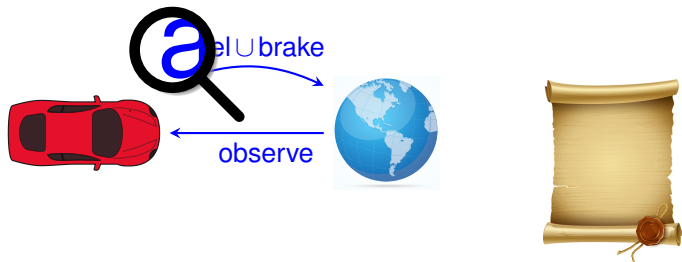
ModelPlex monitors conjunction of all plausible models



Remove incompatible models after contradictory observation



Plan differentiating experiment \leftarrow predictive monitor distinctions



Convergence

Plausible models converge to true model a.s., if possible

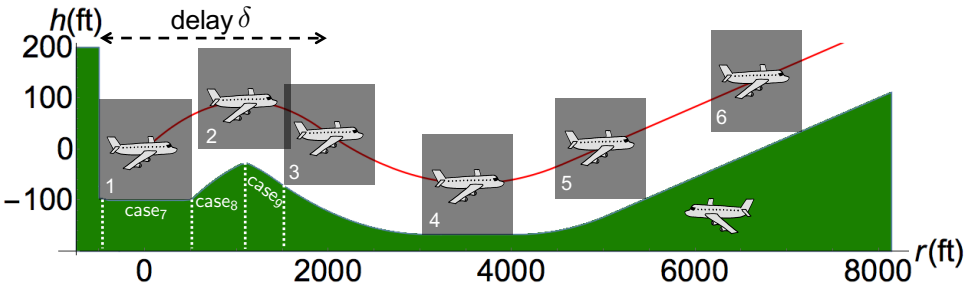


Modify model to fit observations by verification-preserving model update.
Safety proofs reified: modify model + proof tactic to preserve fit + safety



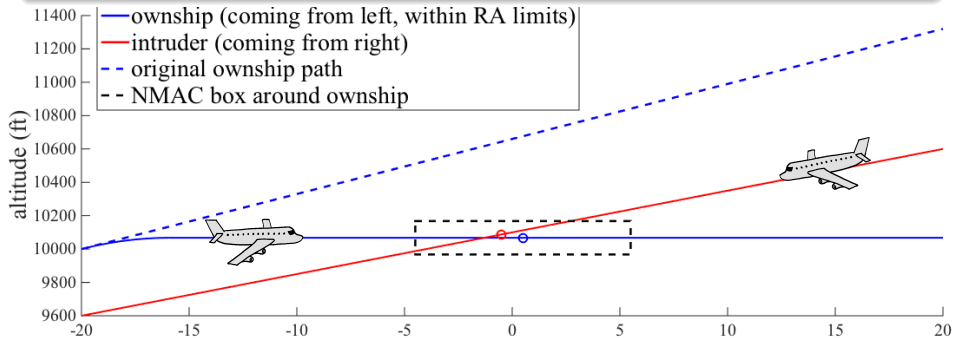
- 1 Autonomous Cyber-Physical Systems
- 2 Foundation: Differential Dynamic Logic
- 3 ModelPlex: Model Safety Transfer
- 4 VeriPhy: Executable Proof Transfer
- 5 Safe Learning in CPSs
- 6 Applications**
 - Airborne Collision Avoidance System
 - Ground Robot Navigation
- 7 Summary

- Developed by the FAA to replace current TCAS in aircraft
- Approximately optimizes Markov Decision Process on a grid
- Advisory from lookup tables with numerous 5D interpolation regions



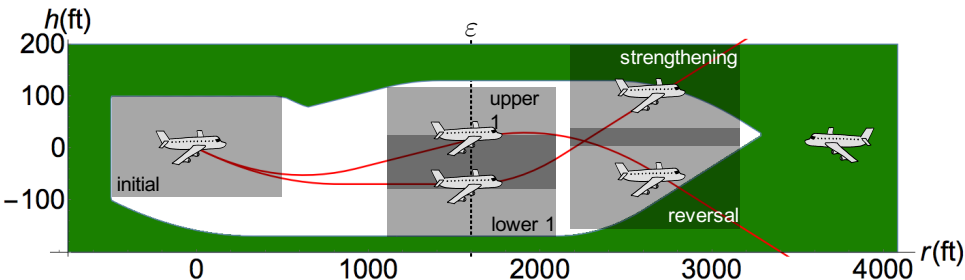
- 1 Identified safe region for each advisory symbolically
- 2 Proved safety for hybrid systems flight model in KeYmaera X

ACAS X table comparison shows safe advisory in 97.7% of the 648,591,384,375 states compared (15,160,434,734 counterexamples).



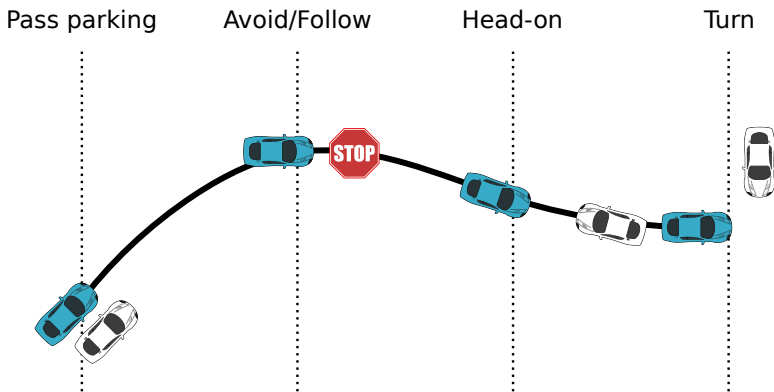
ACAS X issues DNC advisory, which induces collision unless corrected

- Conservative, so too many counterexamples
- Settle for: safe for a little while, with safe future advisory possibility
- Safeable advisory: a subsequent advisory can safely avoid collision



- 1 Identified safeable region for each advisory symbolically
- 2 Proved safety for hybrid systems flight model in KeYmaera X

- Fundamental safety question for ground robot navigation IJRR'17
- When will which control decision avoid obstacles?
- Depends on safety objective, physical capabilities of robot + obstacle



- 1 Identified safe region for each safety notion symbolically
- 2 Proved safety for hybrid systems ground robot model in KeYmaera X

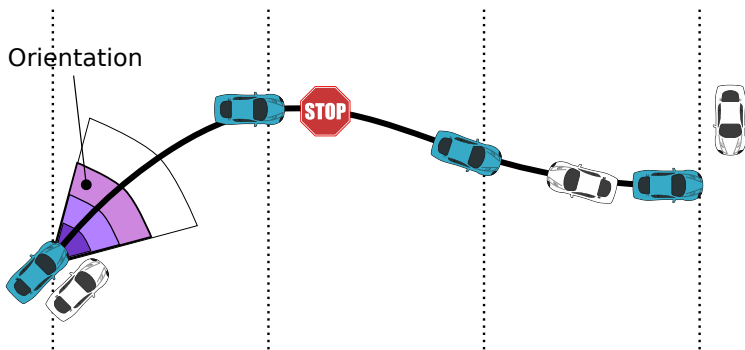
- Fundamental safety question for ground robot navigation IJRR'17
- When will which control decision avoid obstacles?
- Depends on safety objective, physical capabilities of robot + obstacle

Pass parking

Avoid/Follow

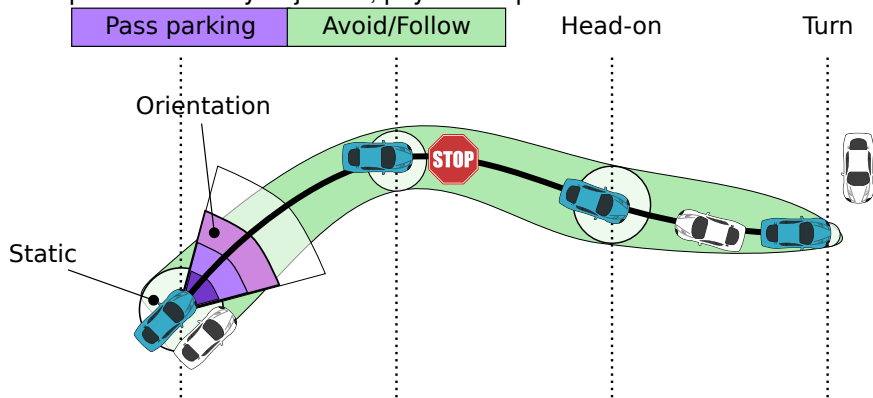
Head-on

Turn



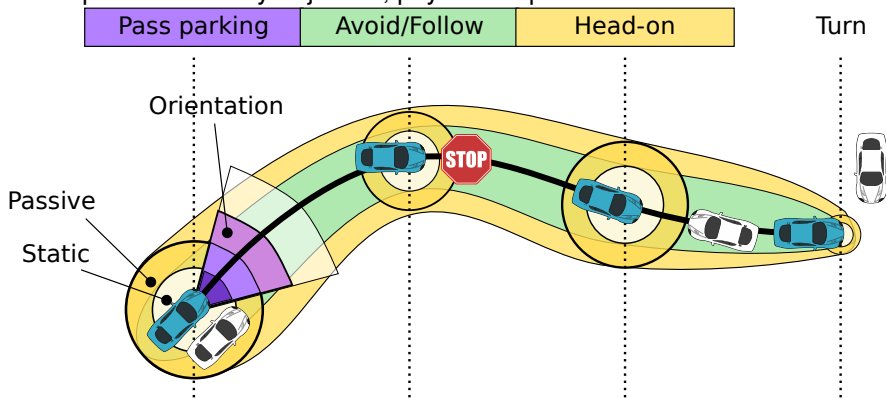
- 1 Identified safe region for each safety notion symbolically
- 2 Proved safety for hybrid systems ground robot model in KeYmaera X

- Fundamental safety question for ground robot navigation IJRR'17
- When will which control decision avoid obstacles?
- Depends on safety objective, physical capabilities of robot + obstacle



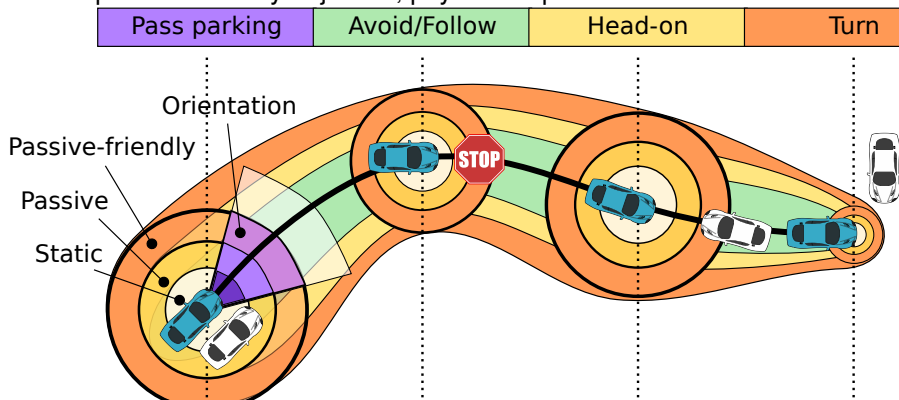
- 1 Identified safe region for each safety notion symbolically
- 2 Proved safety for hybrid systems ground robot model in KeYmaera X

- Fundamental safety question for ground robot navigation
- When will which control decision avoid obstacles?
- Depends on safety objective, physical capabilities of robot + obstacle



- 1 Identified safe region for each safety notion symbolically
- 2 Proved safety for hybrid systems ground robot model in KeYmaera X

- Fundamental safety question for ground robot navigation
- When will which control decision avoid obstacles?
- Depends on safety objective, physical capabilities of robot + obstacle



- 1 Identified safe region for each safety notion symbolically
- 2 Proved safety for hybrid systems ground robot model in KeYmaera X

Safety ▶

Invariant + Safe Control

$$\text{static} \quad \|p - o\|_\infty > \frac{s^2}{2b} + \left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon s\right)$$

$$\text{passive} \quad s \neq 0 \rightarrow \|p - o\|_\infty > \frac{s^2}{2b} + V\frac{s}{b} + \left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(s + V)\right)$$

$$+ \text{ sensor} \quad \|\hat{p} - o\|_\infty > \frac{s^2}{2b} + V\frac{s}{b} + \left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(s + V)\right) + \Delta_p$$

$$+ \text{ disturb.} \quad \|p - o\|_\infty > \frac{s^2}{2b\Delta_a} + V\frac{s}{b\Delta_a} + \left(\frac{A}{b\Delta_a} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(s + V)\right)$$

$$+ \text{ failure} \quad \|\hat{p} - o\|_\infty > \frac{s^2}{2b} + V\frac{s}{b} + \left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(v + V)\right) + \Delta_p + g\Delta$$

$$\text{friendly} \quad \|p - o\|_\infty > \frac{s^2}{2b} + \frac{V^2}{2b_0} + V\left(\frac{s}{b} + \tau\right) + \left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(s + V)\right)$$

⋮



Safety	Invariant	Safe Control
static	$\ p - o\ _\infty > \frac{s^2}{2b}$	$\left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon s\right)$
passive	$s \neq 0 \rightarrow \ p - o\ _\infty > \frac{s^2}{2b}$	$+V\frac{s}{b} + \left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(s+V)\right)$
+ sensor		$+ \Delta_p$
+ disturb.	$\ p - o\ _\infty > \frac{s^2}{2b\Delta_a} + V\frac{s}{b\Delta_a}$	$\left(\frac{A}{b\Delta_a} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(s+V)\right)$
+ failure	$\ \hat{p} - o\ _\infty > \frac{s^2}{2b} + V\frac{s}{b}$	$\left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(v+V)\right) + \Delta_p + g\Delta$
friendly	$\ p - o\ _\infty > \frac{s^2}{2b} + \frac{V^2}{2b_0} + V\left(\frac{s}{b} + \tau\right)$	$\left(\frac{A}{b} + 1\right) \left(\frac{A}{2}\varepsilon^2 + \varepsilon(s+V)\right)$

Question

How to find and justify constraints? Proof!

⋮



- 1 Autonomous Cyber-Physical Systems
- 2 Foundation: Differential Dynamic Logic
- 3 ModelPlex: Model Safety Transfer
- 4 VeriPhy: Executable Proof Transfer
- 5 Safe Learning in CPSs
- 6 Applications
 - Airborne Collision Avoidance System
 - Ground Robot Navigation
- 7 Summary

Logical Systems Lab at Carnegie Mellon University, Computer Science
Yong Kiam Tan, Brandon Bohrer, Nathan Fulton, Sarah Loos, Katherine Cordwell
Stefan Mitsch, Khalil Ghorbal, Jean-Baptiste Jeannin, Andrew Sogokon



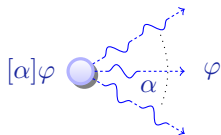
Unterstützt von / Supported by



Alexander von Humboldt
Stiftung / Foundation

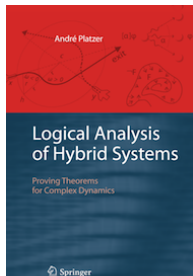
differential dynamic logic

$$dL = DL + HP$$



Logical Triumvirate of Technologies for Transitioning Trustworthiness

- | | |
|-------------------------------------------------------|---------------------------------|
| 1 KeYmaera X: safe action in CPS model | 1 RL optimizes action choice |
| 2 ModelPlex: safe model \rightsquigarrow safe impl | 2 ModelPlex: safe reward for RL |
| 3 VeriPhy: sandbox \rightsquigarrow safe executable | 3 VeriPhy: CPS sandbox for RL |



André Platzer (CMU)

KeYmaera X

KeYmaera X Models Proofs Theme Help

Proof Auto Normalize Step back

Propositional - Hybrid Programs - Differential Equations -

Base case 4 Use case 5 Induction step 6

$\vdash \neg: x \geq 0 \quad \vdash \neg: [x := x + 1; u \{x' = v\}] x \geq 0$

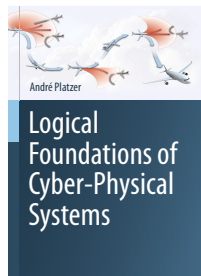
$\vdash \neg: v \geq 0$

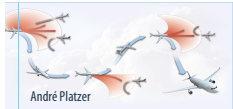
loop $\vdash x \geq 0, v \geq 0 \quad \vdash \neg: [x := x + 1; u \{x' = v\}]^* x \geq 0$

$\vdash \neg: x \geq 0 \wedge v \geq 0 \rightarrow \vdash \neg: [x := x + 1; u \{x' = v \wedge \text{true}\}]^* x \geq 0$

$\vdash \neg: [a; b]P \rightarrow \neg: [a]P \wedge [b]P$

The Logical Path to Autonomous Cyber-Physical Systems





Logical Foundations of Cyber-Physical Systems

Springer



Logical Analysis of Hybrid Systems

Proving Theorems for Complex Dynamics

Springer

Definition (Hybrid program α)

$$x := f(x) \mid ?Q \mid x' = f(x) \& Q \mid \alpha \cup \beta \mid \alpha; \beta \mid \alpha^*$$

Definition (dL Formula P)

$$e \geq \tilde{e} \mid \neg P \mid P \wedge Q \mid \forall x P \mid \exists x P \mid [\alpha]P \mid \langle \alpha \rangle P$$

Discrete
Assign

Test
Condition

Differential
Equation

Nondet.
Choice

Seq.
Compose

Nondet.
Repeat

Definition (Hybrid program α)

$x := f(x) \mid ?Q \mid x' = f(x) \& Q \mid \alpha \cup \beta \mid \alpha; \beta \mid \alpha^*$

Definition (dL Formula P)

$e \geq \tilde{e} \mid \neg P \mid P \wedge Q \mid \forall x P \mid \exists x P \mid [\alpha]P \mid \langle \alpha \rangle P$

All
Reals

Some
Reals

All
Runs

Some
Runs

Definition (Hybrid program semantics)

$([\cdot] : \text{HP} \rightarrow \wp(\mathcal{S} \times \mathcal{S}))$

$$[x := e] = \{(\omega, \nu) : \nu = \omega \text{ except } \nu[x] = \omega[e]\}$$

$$[?Q] = \{(\omega, \omega) : \omega \in [Q]\}$$

$$[x' = f(x)] = \{(\varphi(0), \varphi(r)) : \varphi \models x' = f(x) \text{ for some duration } r\}$$

$$[\alpha \cup \beta] = [\alpha] \cup [\beta]$$

$$[\alpha; \beta] = [\alpha] \circ [\beta]$$

$$[\alpha^*] = [\alpha]^* = \bigcup_{n \in \mathbb{N}} [\alpha^n]$$

compositional semantics

Definition (dL semantics)

$([\cdot] : \text{Fml} \rightarrow \wp(\mathcal{S}))$

$$[e \geq \tilde{e}] = \{\omega : \omega[e] \geq \omega[\tilde{e}]\}$$

$$[\neg P] = [P]^c$$

$$[P \wedge Q] = [P] \cap [Q]$$

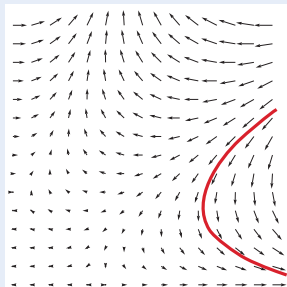
$$[\langle \alpha \rangle P] = [\alpha] \circ [P] = \{\omega : \nu \in [P] \text{ for some } \nu : (\omega, \nu) \in [\alpha]\}$$

$$[[\alpha]P] = [\neg \langle \alpha \rangle \neg P] = \{\omega : \nu \in [P] \text{ for all } \nu : (\omega, \nu) \in [\alpha]\}$$

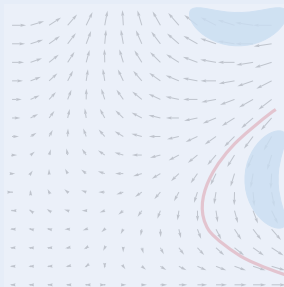
$$[\exists x P] = \{\omega : \omega_x^r \in [P] \text{ for some } r \in \mathbb{R}\}$$

A Differential Invariants for Differential Equations

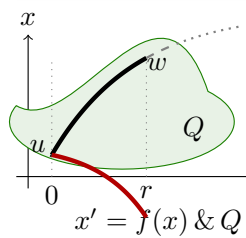
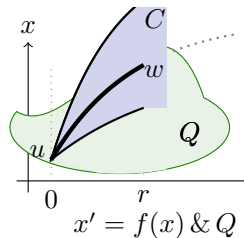
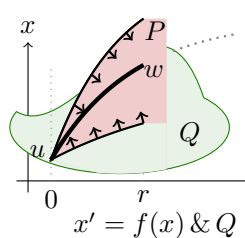
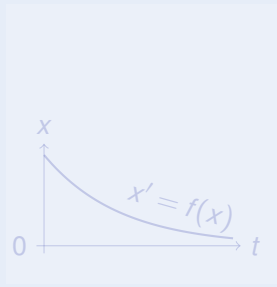
Differential Invariant



Differential Cut

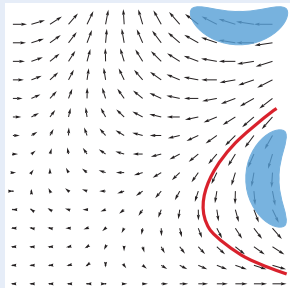


Differential Ghost

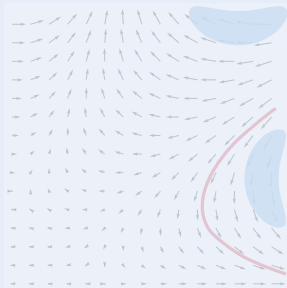


A Differential Invariants for Differential Equations

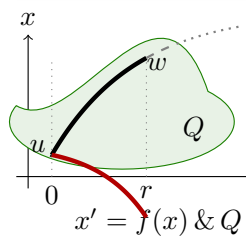
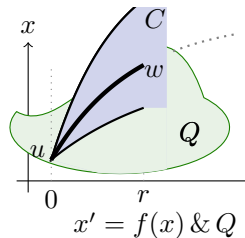
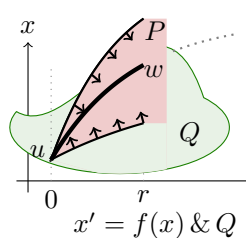
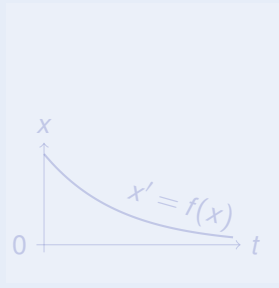
Differential Invariant



Differential Cut

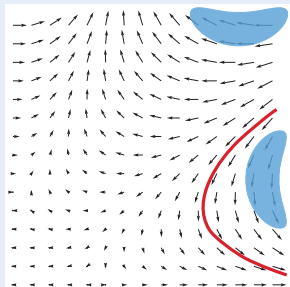


Differential Ghost

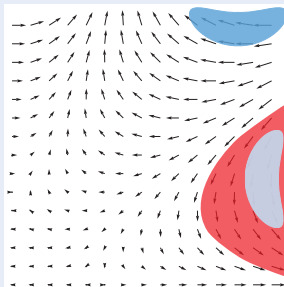


A Differential Invariants for Differential Equations

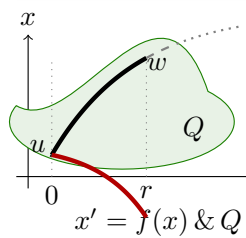
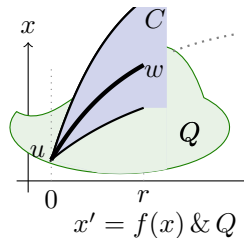
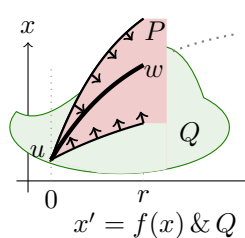
Differential Invariant



Differential Cut

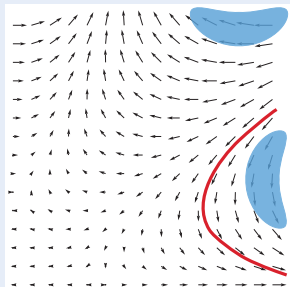


Differential Ghost

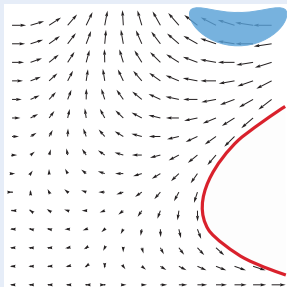


A Differential Invariants for Differential Equations

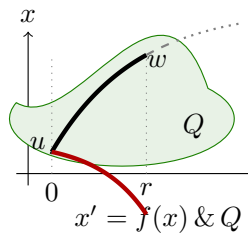
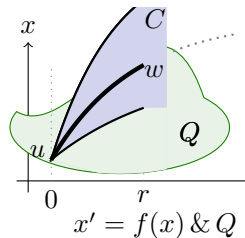
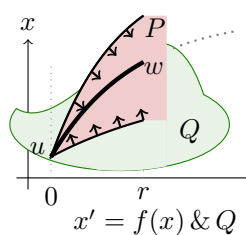
Differential Invariant



Differential Cut

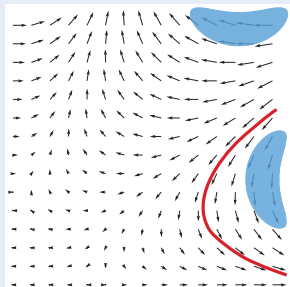


Differential Ghost

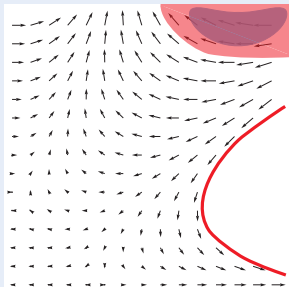


A Differential Invariants for Differential Equations

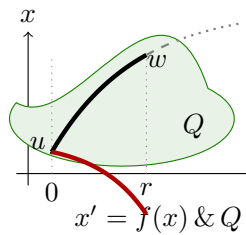
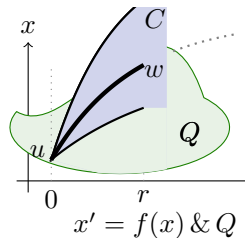
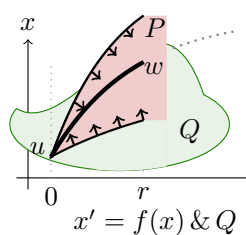
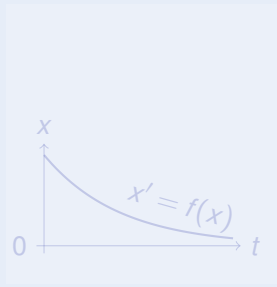
Differential Invariant



Differential Cut

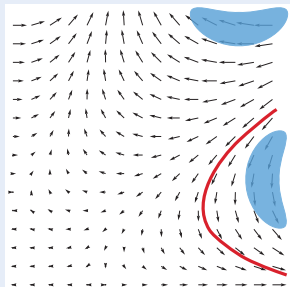


Differential Ghost

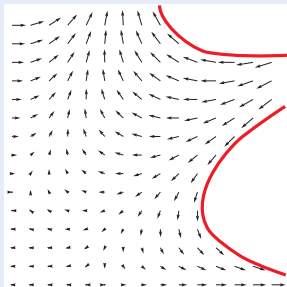


A Differential Invariants for Differential Equations

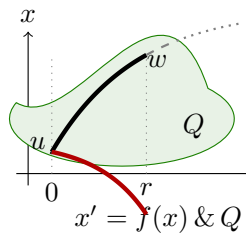
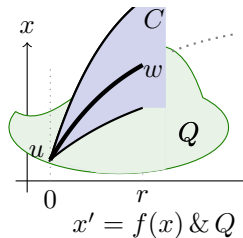
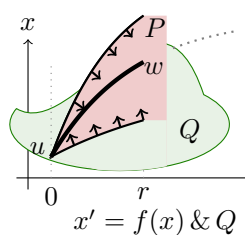
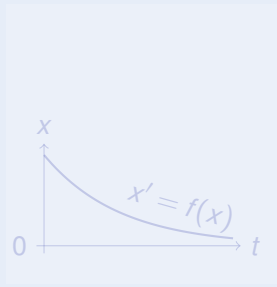
Differential Invariant



Differential Cut

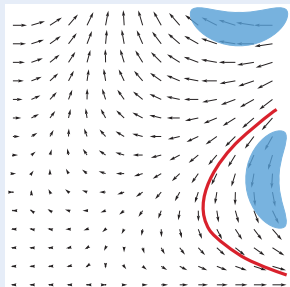


Differential Ghost

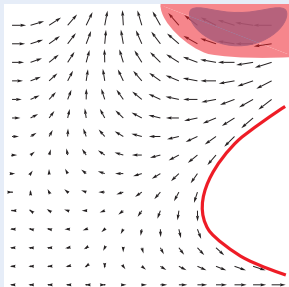


A Differential Invariants for Differential Equations

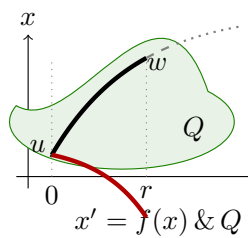
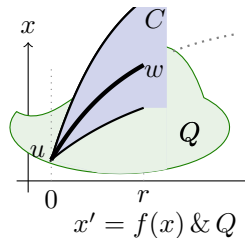
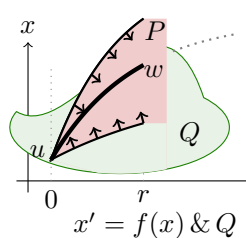
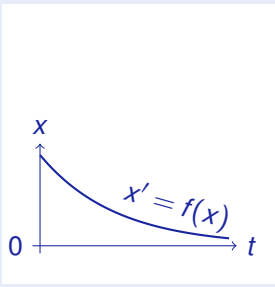
Differential Invariant



Differential Cut

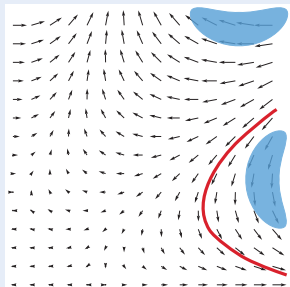


Differential Ghost

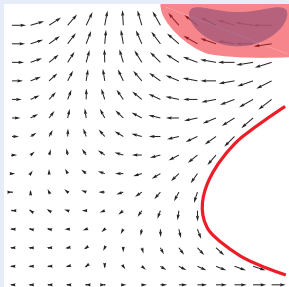


A Differential Invariants for Differential Equations

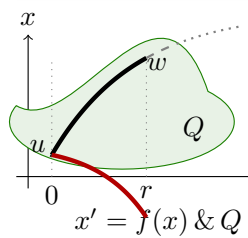
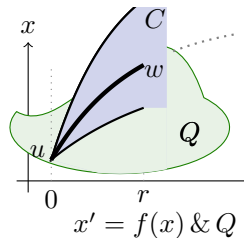
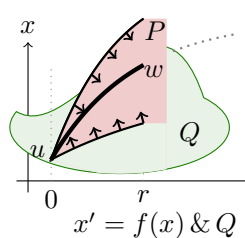
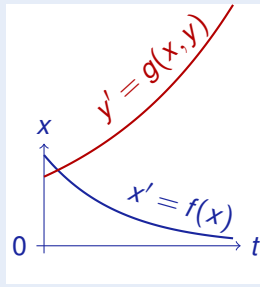
Differential Invariant



Differential Cut

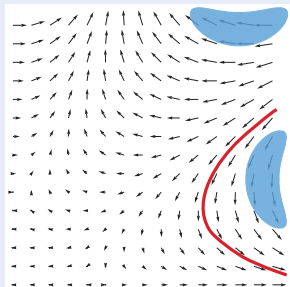


Differential Ghost

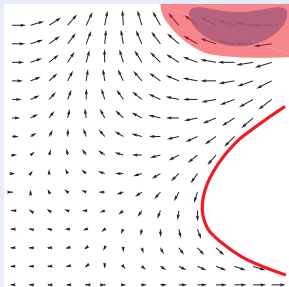


A Differential Invariants for Differential Equations

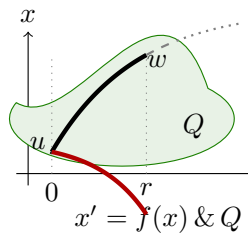
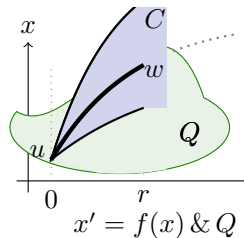
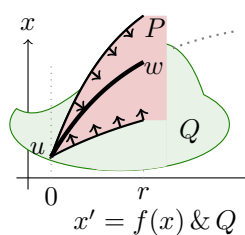
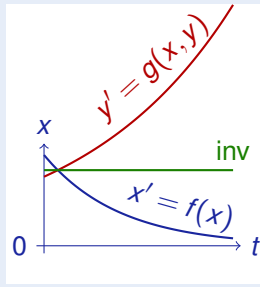
Differential Invariant



Differential Cut



Differential Ghost



Differential Invariants for Differential Equations

Differential Invariant

$$\frac{Q \vdash [x' := f(x)](P)'}{P \vdash [x' = f(x) \& Q]P}$$

Differential Cut

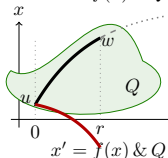
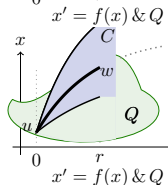
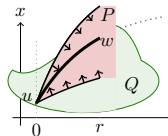
$$\frac{P \vdash [x' = f(x) \& Q]C \quad P \vdash [x' = f(x) \& Q \wedge C]P}{P \vdash [x' = f(x) \& Q]P}$$

Differential Ghost

$$\frac{P \leftrightarrow \exists y G \quad G \vdash [x' = f(x), y' = g(x, y) \& Q]G}{P \vdash [x' = f(x) \& Q]P}$$

deductive power added $DI \prec DI+DC \prec DI+DC+DG$

$$\omega[[e]'] = \sum_x \omega(x') \frac{\partial [[e]]}{\partial x}(\omega)$$



A Differential Invariants for Differential Equations

Differential Invariant

$$\frac{Q \vdash [x' := f(x)](P)'}{P \vdash [x' = f(x) \& Q]P}$$

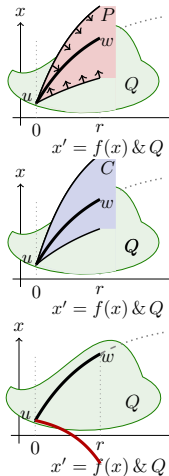
Differential Cut

$$\frac{P \vdash [x' = f(x) \& Q]C \quad P \vdash [x' = f(x) \& Q \wedge C]P}{P \vdash [x' = f(x) \& Q]P}$$

Differential Ghost

$$\frac{P \leftrightarrow \exists y G \quad G \vdash [x' = f(x), y' = g(x, y) \& Q]G}{P \vdash [x' = f(x) \& Q]P}$$

if $g(x, y) = a(x)y + b(x)$, so has long solution!





André Platzer.

Logical Foundations of Cyber-Physical Systems.

Springer, Cham, 2018.

doi:10.1007/978-3-319-63588-0.



André Platzer.

The logical path to autonomous cyber-physical systems.

In David Parker and Verena Wolf, editors, *QEST*, volume 11785 of *LNCS*, pages 25–33. Springer, 2019.

doi:10.1007/978-3-030-30281-8_2.



Nathan Fulton and André Platzer.

Safe reinforcement learning via formal methods: Toward safe control through proof and learning.

In Sheila A. McIlraith and Kilian Q. Weinberger, editors, *AAAI*, pages 6485–6492. AAAI Press, 2018.



Nathan Fulton and André Platzer.

Verifiably safe off-model reinforcement learning.

In Tomas Vojnar and Lijun Zhang, editors, *TACAS, Part I*, volume 11427 of *LNCS*, pages 413–430. Springer, 2019.

[doi:10.1007/978-3-030-17462-0_28](https://doi.org/10.1007/978-3-030-17462-0_28).



André Platzer.

Differential dynamic logic for hybrid systems.

J. Autom. Reas., 41(2):143–189, 2008.

[doi:10.1007/s10817-008-9103-8](https://doi.org/10.1007/s10817-008-9103-8).



André Platzer.

Logics of dynamical systems.

In LICS [17], pages 13–24.

[doi:10.1109/LICS.2012.13](https://doi.org/10.1109/LICS.2012.13).



André Platzer.

A complete uniform substitution calculus for differential dynamic logic.

J. Autom. Reas., 59(2):219–265, 2017.

[doi:10.1007/s10817-016-9385-1](https://doi.org/10.1007/s10817-016-9385-1).



André Platzer.

The complete proof theory of hybrid systems.

In LICS [17], pages 541–550.

[doi:10.1109/LICS.2012.64](https://doi.org/10.1109/LICS.2012.64).



André Platzer and Yong Kiam Tan.

Differential equation axiomatization: The impressive power of differential ghosts.

In Anuj Dawar and Erich Grädel, editors, *LICS*, pages 819–828, New York, 2018. ACM.

[doi:10.1145/3209108.3209147](https://doi.org/10.1145/3209108.3209147).



André Platzer and Yong Kiam Tan.

Differential equation invariance axiomatization.

J. ACM, 67(1):6:1–6:66, 2020.

[doi:10.1145/3380825](https://doi.org/10.1145/3380825).



Stefan Mitsch and André Platzer.

ModelPlex: Verified runtime validation of verified cyber-physical system models.

Form. Methods Syst. Des., 49(1-2):33–74, 2016.

Special issue of selected papers from RV'14.

[doi:10.1007/s10703-016-0241-z](https://doi.org/10.1007/s10703-016-0241-z).



Jean-Baptiste Jeannin, Khalil Ghorbal, Yanni Kouskoulas, Aurora Schmidt, Ryan Gardner, Stefan Mitsch, and André Platzer.

A formally verified hybrid system for safe advisories in the next-generation airborne collision avoidance system.

STTT, 19(6):717–741, 2017.

doi:10.1007/s10009-016-0434-1.



Stefan Mitsch, Khalil Ghorbal, David Vogelbacher, and André Platzer.

Formal verification of obstacle avoidance and navigation of ground robots.

I. J. Robotics Res., 36(12):1312–1340, 2017.

doi:10.1177/0278364917733549.



André Platzer.

Logical Analysis of Hybrid Systems: Proving Theorems for Complex Dynamics.

Springer, Heidelberg, 2010.

doi:10.1007/978-3-642-14509-4.



André Platzer.

Differential-algebraic dynamic logic for differential-algebraic programs.

J. Log. Comput., 20(1):309–352, 2010.

doi:10.1093/logcom/exn070.



André Platzer.

The structure of differential invariants and differential cut elimination.

Log. Meth. Comput. Sci., 8(4:16):1–38, 2012.

doi:10.2168/LMCS-8(4:16)2012.



Logic in Computer Science (LICS), 2012 27th Annual IEEE Symposium on, Los Alamitos, 2012. IEEE.