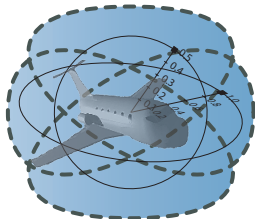


Logic and Compositional Verification of Stochastic Hybrid Systems

André Platzer

Carnegie Mellon University, Pittsburgh, PA





- 1 Motivation
- 2 Stochastic Differential Dynamic Logic $Sd\mathcal{L}$
 - Design
 - Stochastic Differential Equations
 - Syntax
 - Semantics
 - Well-definedness
- 3 Stochastic Differential Dynamic Logic
 - Syntax
 - Semantics
 - Well-definedness
- 4 Proof Calculus for Stochastic Hybrid Systems
 - Compositional Proof Calculus
 - Soundness
- 5 Conclusions

Q: I want to verify trains

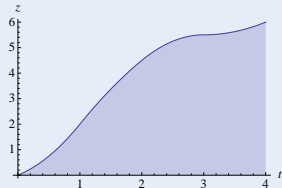
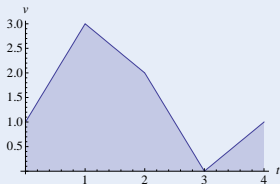
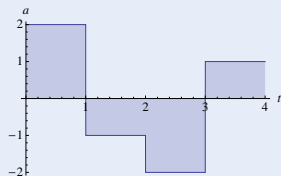
Challenge



Q: I want to verify trains A: Hybrid systems

Challenge (Hybrid Systems)

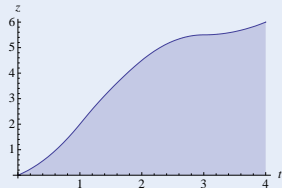
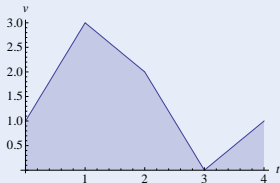
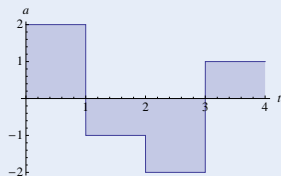
- Continuous dynamics (differential equations)
- Discrete dynamics (control decisions)



Q: I want to verify trains A: Hybrid systems Q: But there's uncertainties!

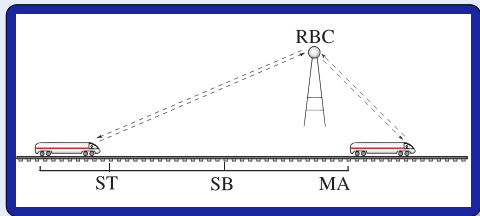
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Q: I want to verify uncertain trains

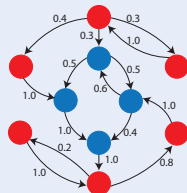
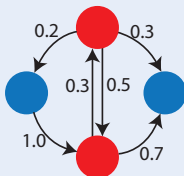
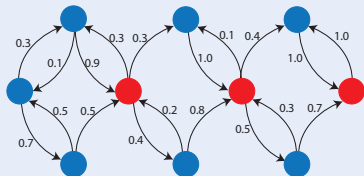
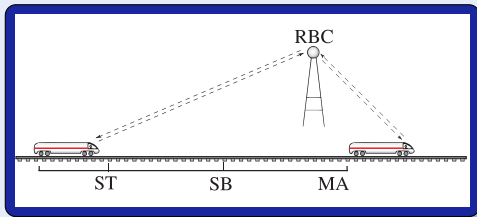
Challenge



Q: I want to verify uncertain trains A: Markov chains

Challenge (Probabilistic Systems)

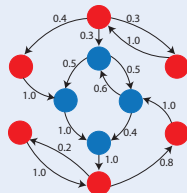
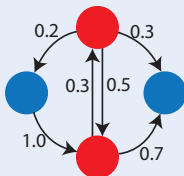
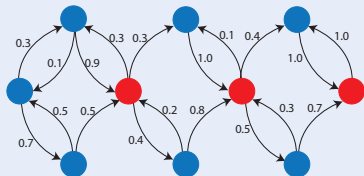
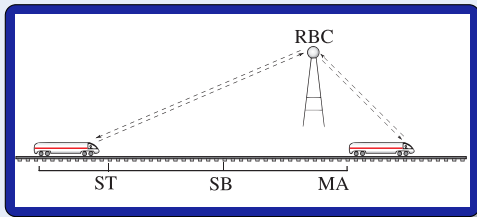
- Directed graph (Countable state space)
- Weighted edges (Transition probabilities)



Q: I want to verify uncertain trains A: Markov chains Q: But trains move!

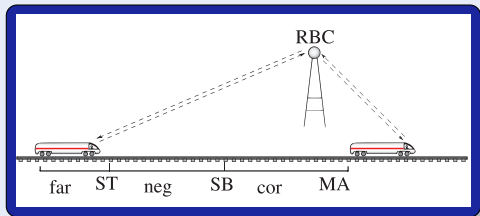
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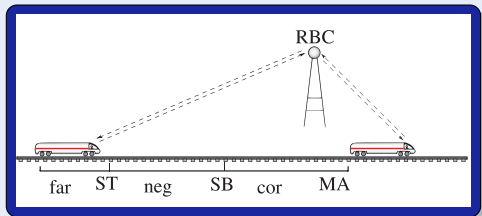
Challenge



Q: I want to verify uncertain trains A: Stochastic hybrid systems

Challenge (Stochastic Hybrid Systems)

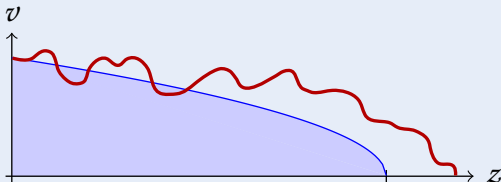
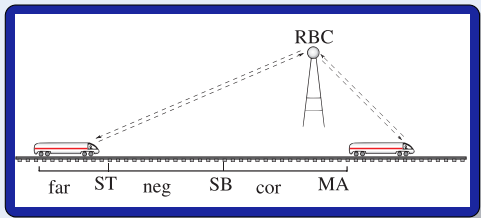
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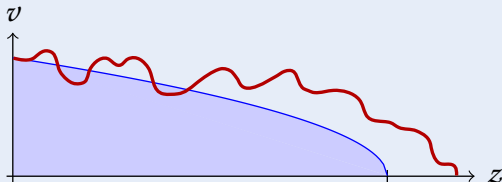
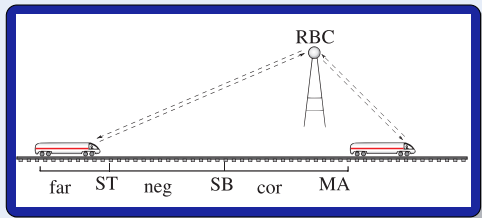
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- Discrete stochastic (lossy communication)
- Continuous stochastic (wind, track)

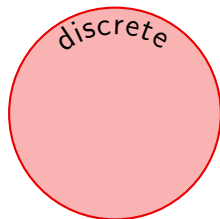


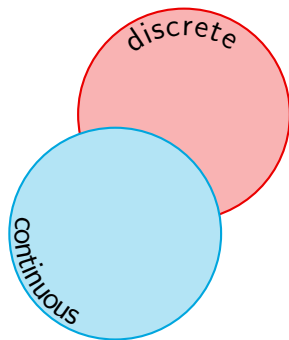
Q: I want to verify uncertain trains A: Stochastic hybrid systems Q: How?

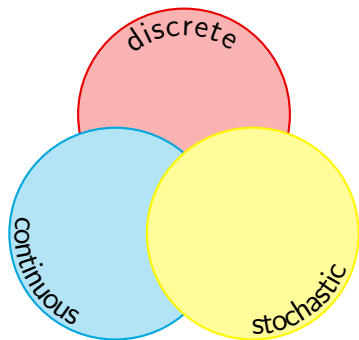
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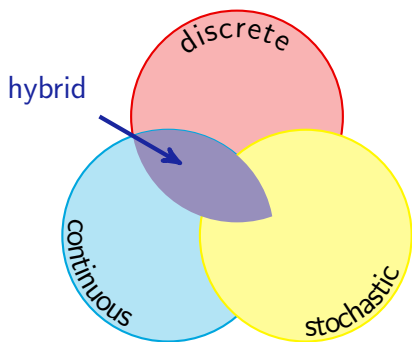
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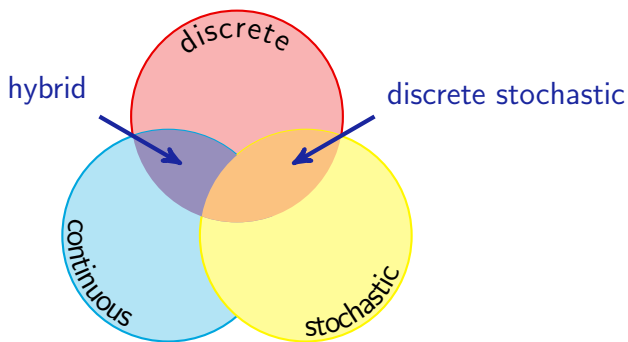


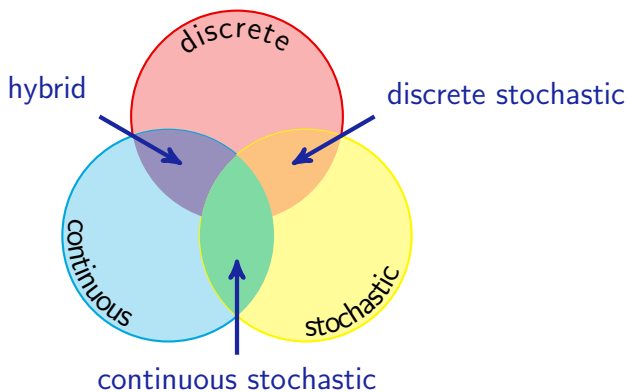


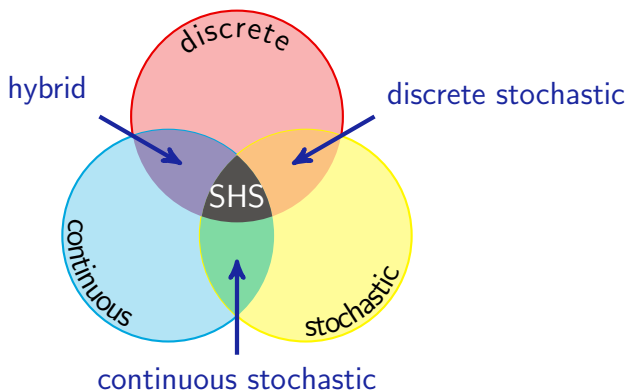












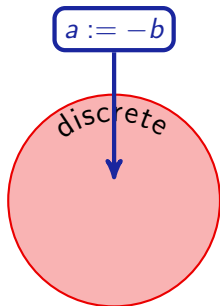


- 1 System model and semantics for stochastic hybrid systems: SHP
- 2 Prove semantic processes are adapted and a.s. càdlàg
- 3 Prove natural process stopping times are Markov times
- 4 Specification and verification logic: $Sd\mathcal{L}$
- 5 Prove measurability of $Sd\mathcal{L}$ semantics \Rightarrow probabilities well-defined
- 6 Proof rules for $Sd\mathcal{L}$
- 7 Sound Dynkin use of infinitesimal generators of SDEs
- 8 First compositional verification for stochastic hybrid systems
- 9 Logical foundation for analysis of stochastic hybrid systems



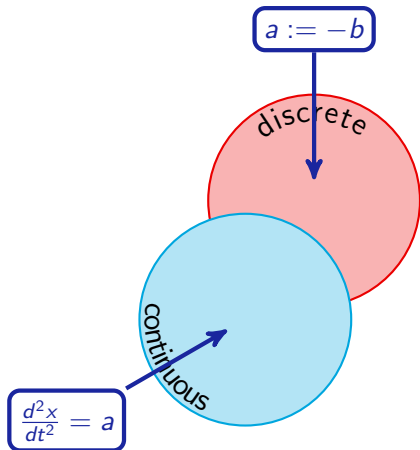
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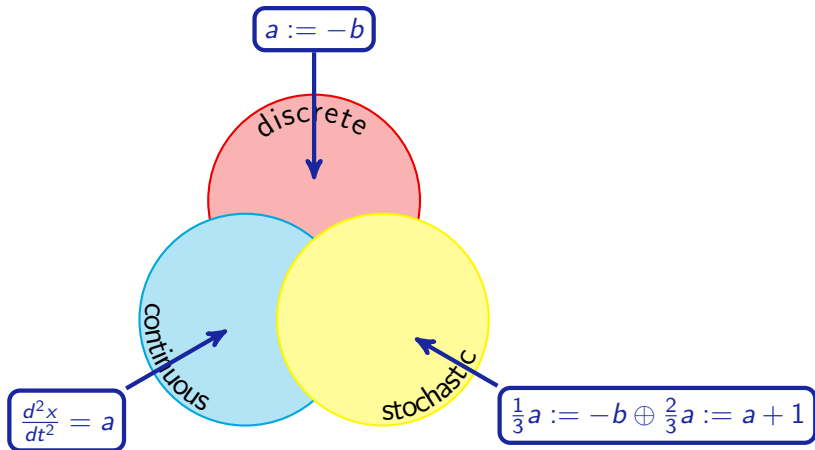
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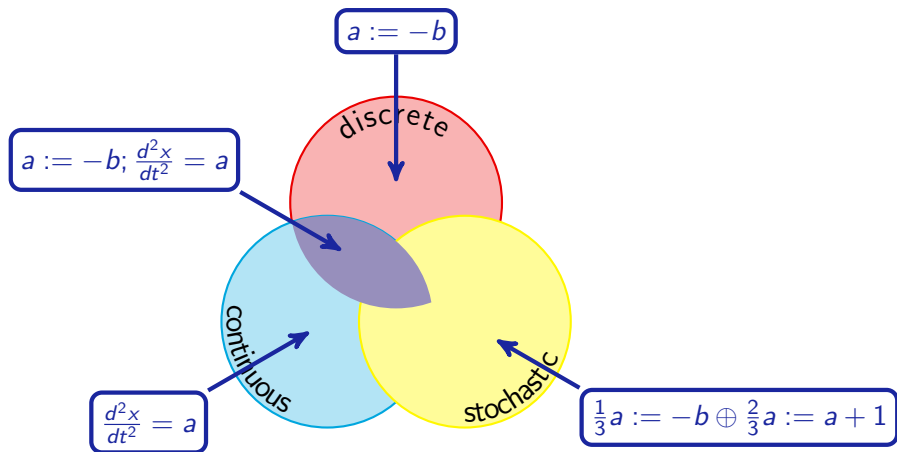
Model for Stochastic Hybrid Systems





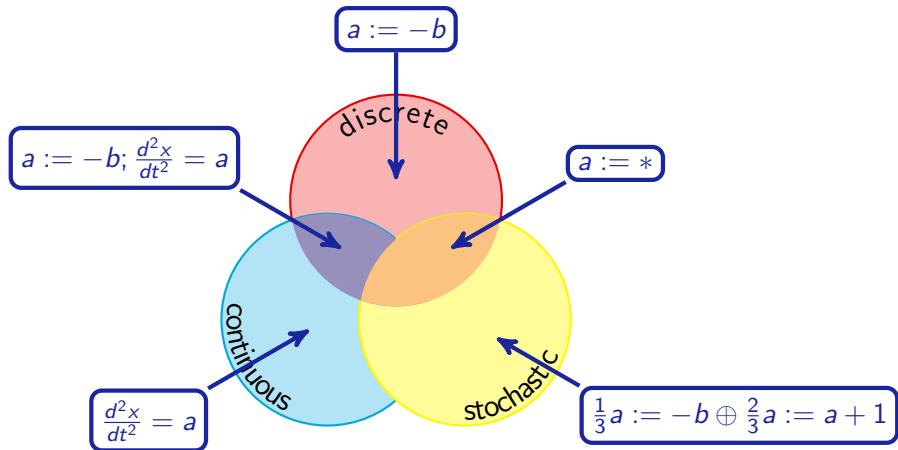


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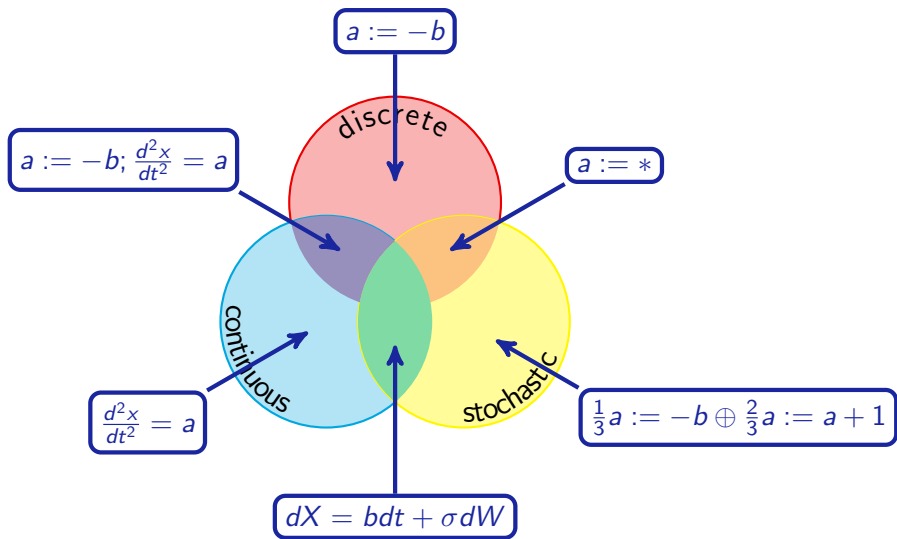


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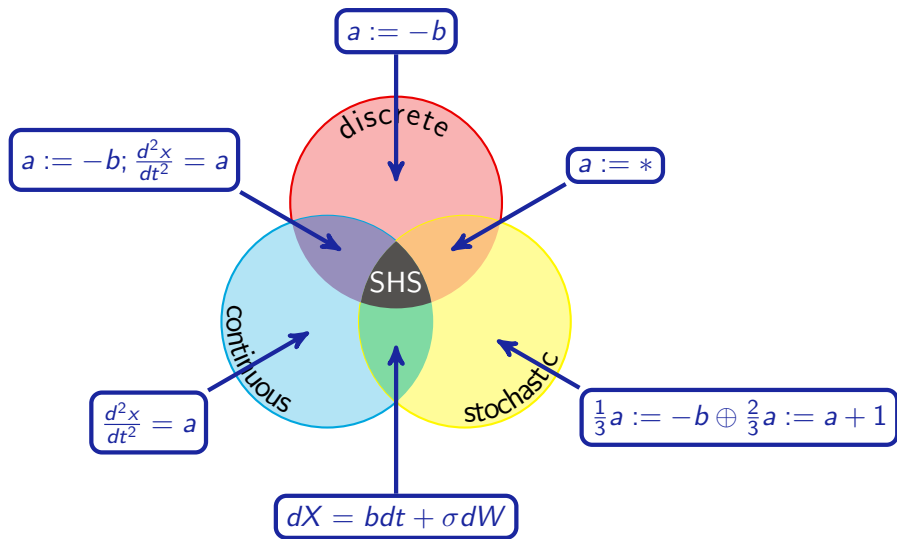


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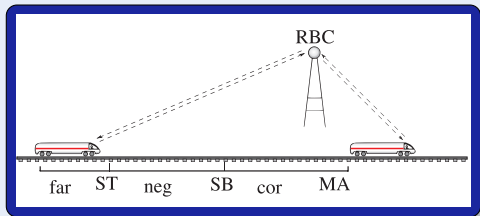


Model for Stochastic Hybrid Systems



Q: How to model stochastic hybrid systems

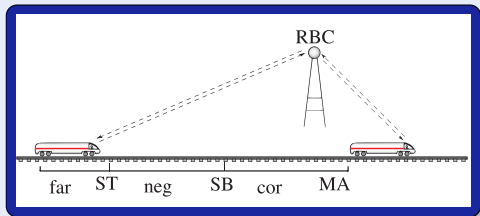
Model (Stochastic Hybrid Systems)



Q: How to model stochastic hybrid systems

Model (Stochastic Hybrid Systems)

- Discrete dynamics
(control decisions)
 $a := -b$
- Continuous dynamics
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- Stochastic dynamics
(structural)



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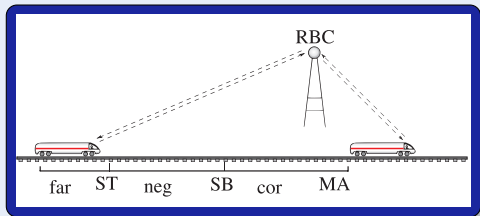
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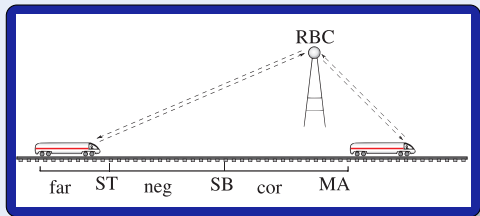
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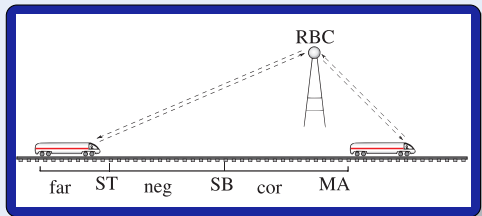
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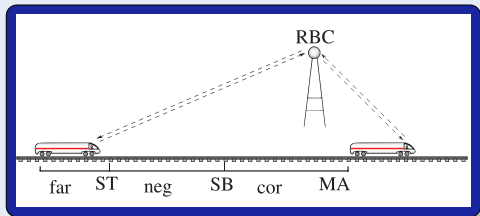
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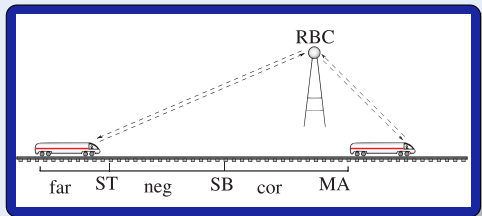
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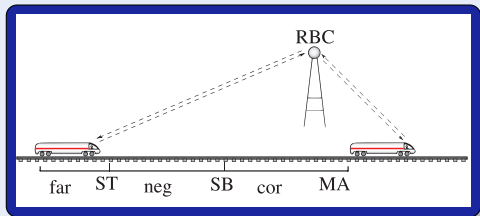
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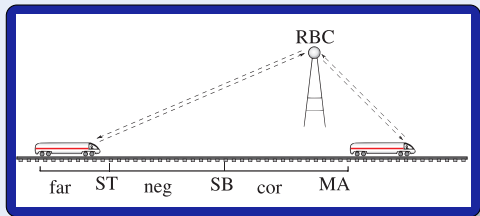
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Q: How to model stochastic hybrid systems A: Stochastic Hybrid Programs

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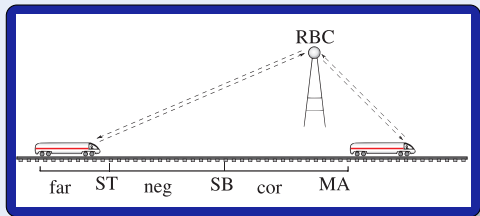
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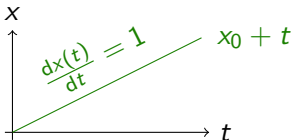




Stochastic Differential Equations (SDE)

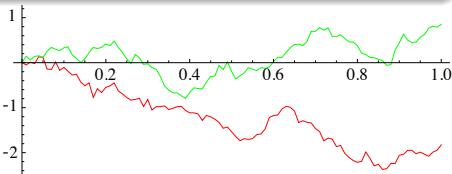
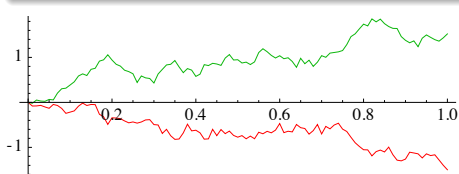
Definition (Ordinary differential equation (ODE))

$$\frac{dx(t)}{dt} = b(x(t)) \quad x(0) = x_0$$



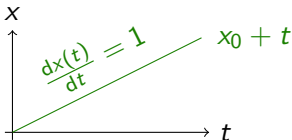
Definition (Itô stochastic differential equation (SDE))

$$dX_t = b(X_t)dt + \sigma(X_t)dW_t \quad X_0 = Z$$



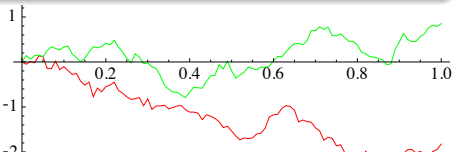
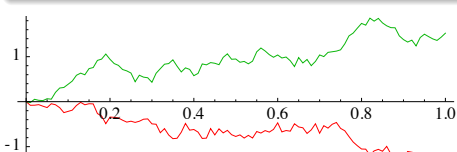
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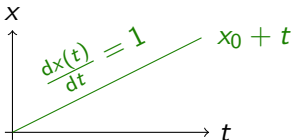
$$X_s = Z + \int_0^s dX_t = Z + \int_0^s b(X_t)dt + \int_0^s \sigma(X_t)dW_t$$



Stochastic Differential Equations (SDE)

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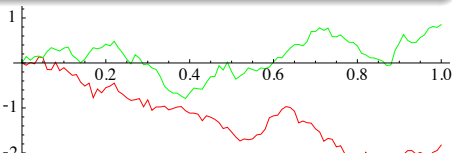
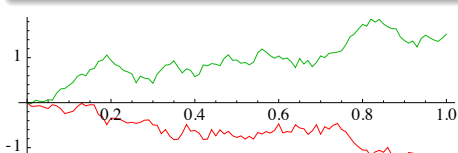
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Calculus

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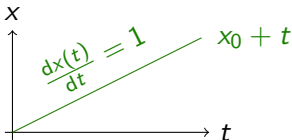
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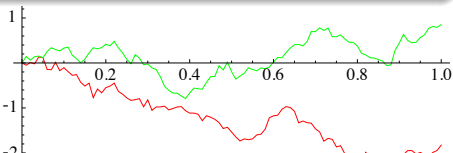
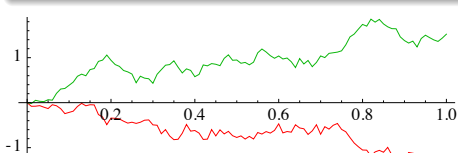
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Definition (Brownian motion W) \Rightarrow end of calculus)

① $W_0 = 0$ (start at 0)

② W_t almost surely continuous

③ $W_t - W_s \sim \mathcal{N}(0, t - s)$ (independent normal increments)

\Rightarrow a.s. continuous everywhere but nowhere differentiable

\Rightarrow a.s. unbounded variation, \notin FV, nonmonotonic on every interval



Brownian Motion is Extremely Complex

Definition (Brownian motion W)

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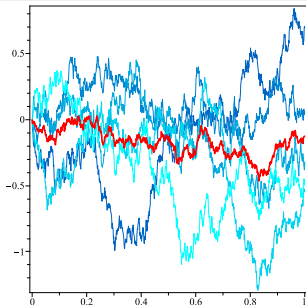
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Definition (Stochastic hybrid program α)

$x := \theta$	(assignment)	} jump & test
$x := *$	(random assignment)	
$?H$	(conditional execution)	
$dx = bdt + \sigma dW \ \& \ H$	(SDE)	} algebra
$\alpha; \beta$	(seq. composition)	
$\lambda\alpha \oplus \nu\beta$	(convex combination)	
α^*	(nondet. repetition)	



What is the Semantics of a Stochastic Hybrid Program?

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- Idea: Start at initial value described by random variable $Z : \Omega \rightarrow \mathbb{R}^d$



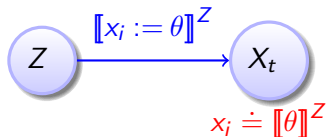
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- When does a stochastic process stop?
- Semantics of program α includes stopping time generator
 $\langle \alpha \rangle : (\Omega \rightarrow \mathbb{R}^d) \rightarrow (\Omega \rightarrow \mathbb{R})$ giving stopping time
 $\langle \alpha \rangle^Z : \Omega \rightarrow \mathbb{R}$ for each Z

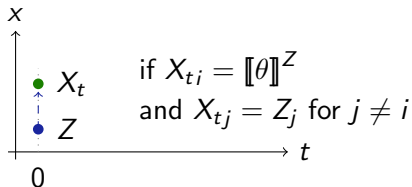


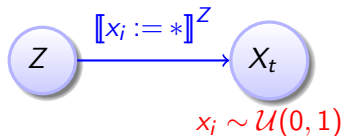
Definition (Stochastic hybrid program α : process semantics



$$[[x_i := \theta]]^Z = \hat{Y} \quad Y(\omega)_i = [[\theta]]^{Z(\omega)} \text{ and } Y_j = Z_j \text{ (for } j \neq i)$$

$$(\downarrow x_i := \theta)^Z = 0$$

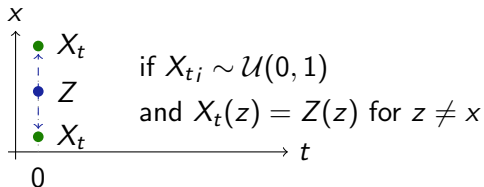


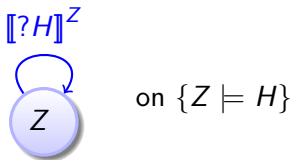


Definition (Stochastic hybrid program α : process semantics ▶▶)

$$\llbracket x_i := * \rrbracket^Z = \hat{U} \quad U_i \sim \mathcal{U}(0, 1) \text{ i.i.d. } \mathcal{F}_0\text{-measurable}$$

$$\langle x_i := * \rangle^Z = 0$$

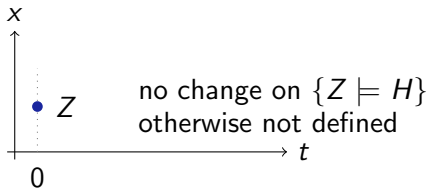


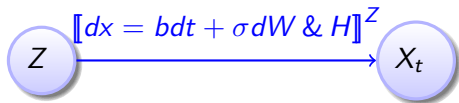


Definition (Stochastic hybrid program α : process semantics ▶▶)

$$[[?H]]^Z = \hat{Z} \quad \text{on the event } \{Z \models H\}$$

$$([?H])^Z = 0$$

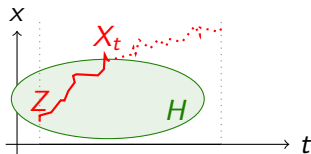




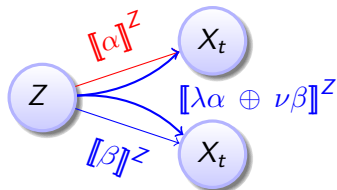
Definition (Stochastic hybrid program α : process semantics ▶▶)

$\llbracket dx = bdt + \sigma dW \ \& \ H \rrbracket^Z$ solves $dX = \llbracket b \rrbracket^X dt + \llbracket \sigma \rrbracket^X dB_t, X_0 = Z$

$(dx = bdt + \sigma dW \ \& \ H)^Z = \inf\{t \geq 0 : X \notin H\}$



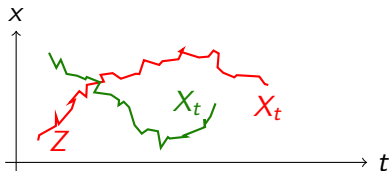
$dx = bdt + \sigma dW \ \& \ H$

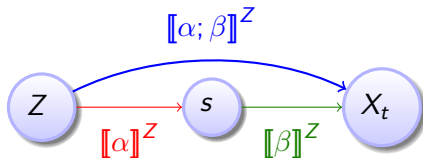


Definition (Stochastic hybrid program α : process semantics)

$$[[\lambda\alpha \oplus \nu\beta]]^Z = \mathcal{I}_{U \leq \lambda} [[\alpha]]^Z + \mathcal{I}_{U > \lambda} [[\beta]]^Z = \begin{cases} [[\alpha]]^Z & \text{on event } \{U \leq \lambda\} \\ [[\beta]]^Z & \text{on event } \{U > \lambda\} \end{cases}$$

$$(|\lambda\alpha \oplus \nu\beta|)^Z = \mathcal{I}_{U \leq \lambda} (|\alpha|)^Z + \mathcal{I}_{U > \lambda} (|\beta|)^Z \text{ with i.i.d. } U \sim \mathcal{U}(0, 1), \mathcal{F}_0\text{-meas}$$

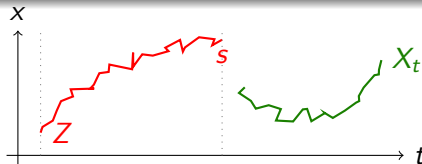




Definition (Stochastic hybrid program α : process semantics)

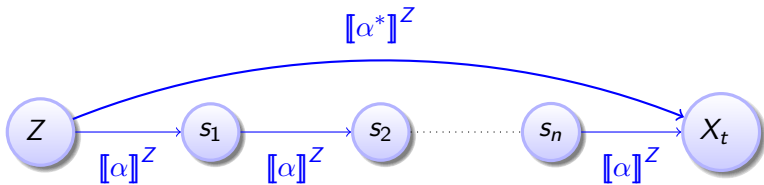
$$[[\alpha; \beta]]_t^Z = \begin{cases} [[\alpha]]_t^Z & \text{on event } \{t < (\alpha)^Z\} \\ [[\beta]]_{t - (\alpha)^Z}^{[[\alpha]]_{(\alpha)^Z}^Z} & \text{on event } \{t \geq (\alpha)^Z\} \end{cases}$$

$$(\alpha; \beta)^Z = (\alpha)^Z + (\beta)^{[[\alpha]]_{(\alpha)^Z}^Z}$$





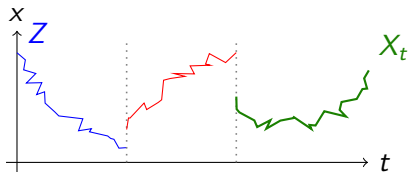
Stochastic Hybrid Program: Process Semantics



Definition (Stochastic hybrid program α : process semantics)

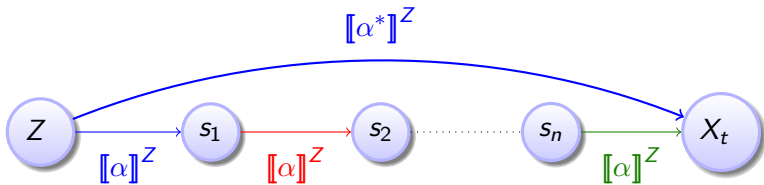
$$[[\alpha^*]]_t^Z = [[\alpha^n]]_t^Z \text{ on event } \{([\alpha^n])^Z > t\}$$

$$([\alpha^*])^Z = \lim_{n \rightarrow \infty} ([\alpha^n])^Z$$





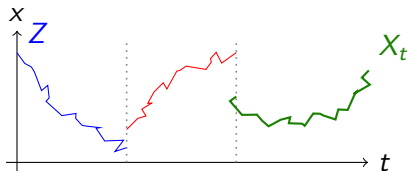
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Definition (Stochastic hybrid program α : process semantics)

$$[[\alpha^*]]_t^Z = [[\alpha^n]]_t^Z \text{ on event } \{([\alpha^n])^Z > t\}$$

$$([\alpha^*])^Z = \lim_{n \rightarrow \infty} ([\alpha^n])^Z \quad \text{monotone!}$$



Theorem

- 1 $[[\alpha]]^Z$ is a.s. càdlàg and adapted
(to completed filtration (\mathcal{F}_t) generated by $Z, (W_s)_{s \leq t}, U$)
 - 2 $(\lfloor \alpha \rfloor)^Z$ is a Markov time / stopping time
(i.e., $\{(\lfloor \alpha \rfloor)^Z \leq t\} \in \mathcal{F}_t$)
- \Rightarrow End value $[[\alpha]]_{(\lfloor \alpha \rfloor)^Z}^Z$ is $\mathcal{F}_{(\lfloor \alpha \rfloor)^Z}$ -measurable.



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Definition (SdL term f)

F	(primitive measurable function, e.g., characteristic \mathcal{I}_A)
$\lambda f + \nu g$	(linear term)
Bf	(scalar term for boolean term B)
$\langle \alpha \rangle f$	(reachable)

Definition (SdL formula ϕ)

$$\phi ::= f \leq g \mid f = g$$

- Semantics of classical logics maps interpretations to truth-values.

What is the Semantics of Sd \mathcal{L} ?

- Semantics of classical logics maps interpretations to truth-values.
- This does not work for Sd \mathcal{L} , because state evolution of α in $\langle \alpha \rangle f$ is stochastic.

- Semantics of classical logics maps interpretations to truth-values.
- This does not work for SdL, because state evolution of α in $\langle \alpha \rangle f$ is stochastic.
- Semantics of SdL is stochastic.
- Semantics of SdL is a random variable generator
 $\llbracket f \rrbracket : (\Omega \rightarrow \mathbb{R}^d) \rightarrow (\Omega \rightarrow \mathbb{R})$ giving a random variable
 $\llbracket f \rrbracket^Z : \Omega \rightarrow \mathbb{R}$ for each initial state random variable Z

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$$\llbracket \langle \alpha \rangle f \rrbracket^Z = \sup \{ \llbracket f \rrbracket^{\llbracket \alpha \rrbracket_t^Z} : 0 \leq t \leq \langle \alpha \rangle^Z \}$$

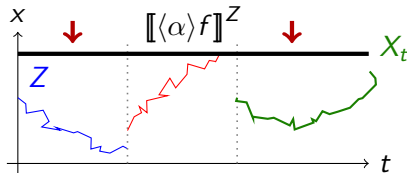
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Theorem (Measurable)

$\llbracket f \rrbracket^Z$ is a random variable (i.e., measurable) for any random variable Z and Sd \mathcal{L} term f .

Theorem (Measurable)

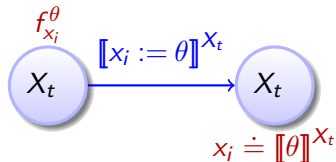
$\llbracket f \rrbracket^Z$ is a random variable (i.e., measurable) for any random variable Z and Sd \mathcal{L} term f .

Corollary (Pushforward measure well-defined for Borel-measurable S)

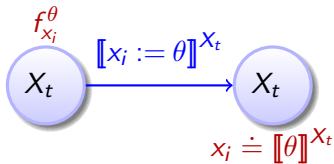
$$S \mapsto P(\llbracket f \rrbracket^Z)^{-1}(S) = P(\{\omega \in \Omega : \llbracket f \rrbracket^Z(\omega) \in S\}) = P(\llbracket f \rrbracket^Z \in S)$$

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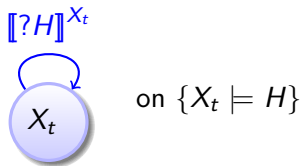
$$\langle x_i := \theta \rangle f = f_{x_i}^\theta$$



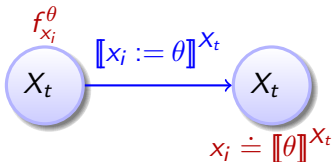
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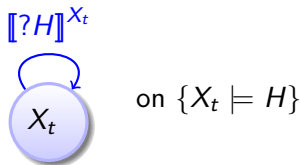
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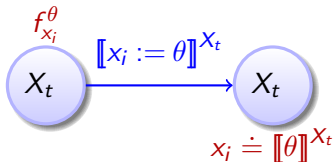


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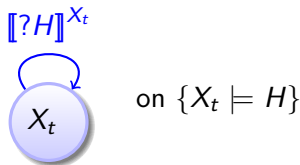


$$\langle \alpha \rangle (\lambda f) = \lambda \langle \alpha \rangle f$$

$$\langle x_i := \theta \rangle f = f_{x_i}^\theta$$



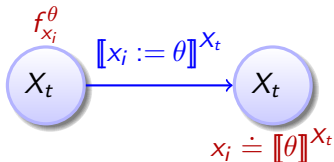
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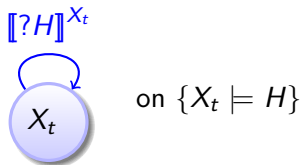
$$\langle \alpha \rangle (\lambda f) = \lambda \langle \alpha \rangle f$$

$$\langle \alpha \rangle (\lambda f + \nu g) \leq \lambda \langle \alpha \rangle f + \nu \langle \alpha \rangle g$$

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$$\langle ?H \rangle f = Hf$$



$$\langle \alpha \rangle (\lambda f) = \lambda \langle \alpha \rangle f$$

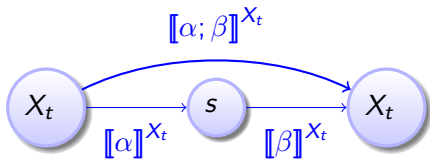
$$\langle \alpha \rangle (\lambda f + \nu g) \leq \lambda \langle \alpha \rangle f + \nu \langle \alpha \rangle g$$

$$f \leq g \models \langle \alpha \rangle f \leq \langle \alpha \rangle g$$

$$\langle \alpha; \beta \rangle f \leq \langle \alpha \rangle (f \sqcup \langle \beta \rangle f)$$

$$f \leq \langle \beta \rangle f \models$$

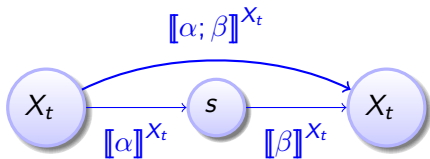
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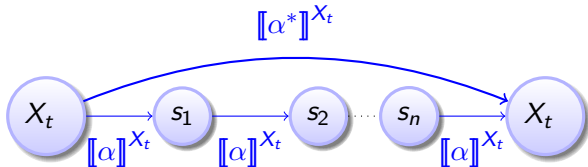
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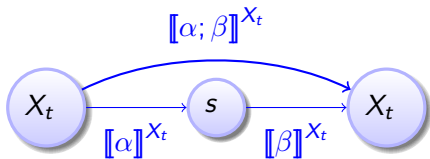
$$\langle \alpha \rangle f \leq f \models \langle \alpha^* \rangle f \leq f$$



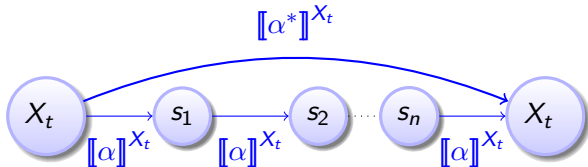
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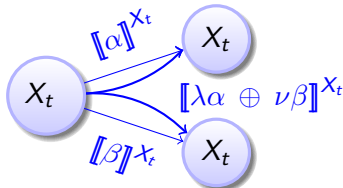
$$\langle \alpha; \beta \rangle f \leq \langle \alpha \rangle \langle \beta \rangle f$$



$$\langle \alpha \rangle f \leq f \models \langle \alpha^* \rangle f \leq f$$



$$\begin{aligned} &P(\langle \lambda \alpha \oplus \nu \beta \rangle f \in S) \\ &= \lambda P(\langle \alpha \rangle f \in S) \\ &+ \nu P(\langle \beta \rangle f \in S) \end{aligned}$$



Theorem (Soundness)

SdL calculus is sound.

- 1 *Rules are globally sound pathwise, i.e., $f_i \leq g_i \models f \leq g$ holds for each initial Z pathwise for each $\omega \in \Omega$*
- 2 *$\langle \oplus \rangle$ is sound in distribution*

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- 2 $\langle \oplus \rangle$ is sound in distribution

Theorem (Soundness for SDE)

Let $\lambda > 0$, $f \in C^2(\mathbb{R}^d, \mathbb{R})$ compact support on H (e.g., H bounded)

$$\frac{\langle \alpha \rangle (H \rightarrow f) \leq \lambda p \quad H \rightarrow f \geq 0 \quad H \rightarrow Lf \leq 0}{P(\langle \alpha \rangle \langle dx = bdt + \sigma dW \ \& \ H \rangle f \geq \lambda) \leq p} \quad \text{sound}$$

$$\frac{\langle \alpha \rangle (H \rightarrow f) \leq \lambda p \quad H \rightarrow f \geq 0 \quad H \rightarrow Lf \leq 0}{P(\langle \alpha \rangle \langle dx = bdt + \sigma dW \ \& \ H \rangle f \geq \lambda) \leq p}$$

$$\langle ?x^2 + y^2 \leq \frac{1}{3} \rangle (H \rightarrow f) = \left(H \rightarrow x^2 + y^2 \leq \frac{1}{3} \right) (x^2 + y^2) \leq 1 * \frac{1}{3}$$

$$f \equiv x^2 + y^2 \geq 0 \quad \text{with} \quad H \equiv x^2 + y^2 < 10$$

$$Lf = \frac{1}{2} \left(-x \frac{\partial f}{\partial x} - y \frac{\partial f}{\partial y} + y^2 \frac{\partial^2 f}{\partial x^2} - 2xy \frac{\partial^2 f}{\partial x \partial y} + x^2 \frac{\partial^2 f}{\partial y^2} \right) \leq 0$$

$$\frac{P(\langle ?x^2 + y^2 \leq \frac{1}{3} \rangle; dx = -\frac{x}{2}dt - ydW, dy = -\frac{y}{2}dt + xdW \ \& \ H \rangle x^2 + y^2 \geq 1)}{1}$$

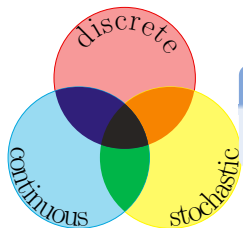
$$\leq \quad (\text{by } \langle ; \rangle')$$

$$P(\langle ?x^2 + y^2 \leq \frac{1}{3} \rangle \langle dx = -\frac{x}{2}dt - ydW, dy = -\frac{y}{2}dt + xdW \ \& \ H \rangle x^2 + y^2 \geq 1)$$

$$\leq \frac{1}{3}$$

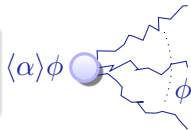


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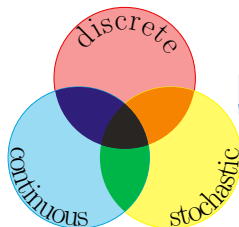


stochastic differential dynamic logic

$$\text{SdL} = \text{DL}_{\text{arithmetic}} + \text{SHP}$$

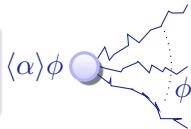


- Stochastic hybrid systems
- Compositional system model & semantics
- Logic for stochastic hybrid systems
- Well-definedness & measurability
- Stochastics accessible in logic
- Compositional proof rules
- Stochastic calculus & symbolic logic

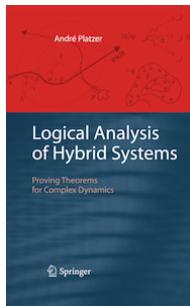


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- Stochastic hybrid systems
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- Extend study of stochastic effects in hybrid systems
- Structural properties of differential invariants
- Computing differential invariants and AI
- Heterogeneity in verification