

Neural Continuous-Time Supermartingale Certificates

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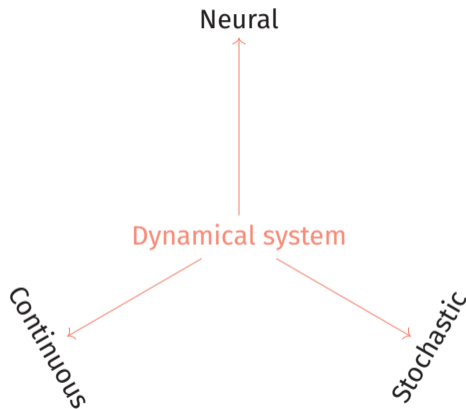
Anna Lukina



What Is This Talk About?

Can we certify safety and/or persistence (\approx stability) of a dynamical system that is:

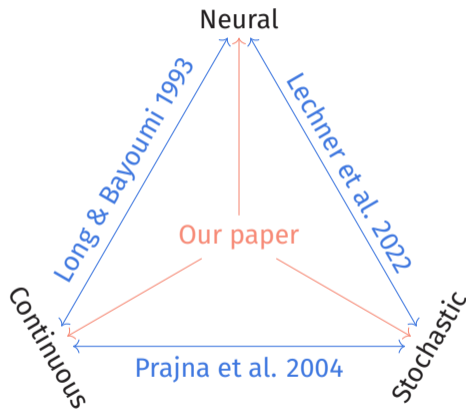
- *neural-controlled*,
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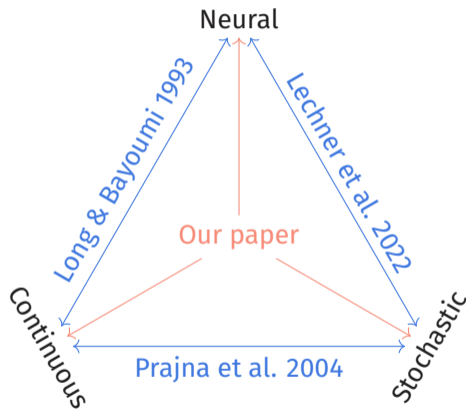


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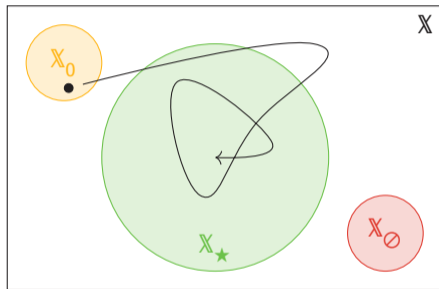
As a bonus, we also allow non-stationarity!



What Exactly Do We Certify?

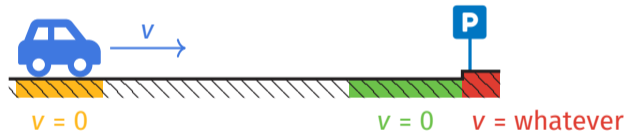
For any state from the **subspace of initial states** X_0 :

- *Reach-avoidance* — the system is able to reach the **subspace of target states** X_\star without entering the **unsafe states** X_\emptyset ;
- *Staying* — at some point the system stays in the **target states** X_\star .



We are interested in formal guarantees of either property (or both).

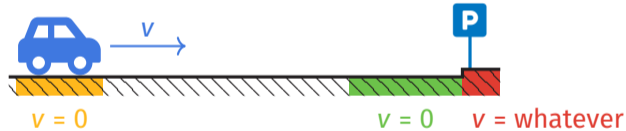
Deterministic Example



Deterministic Example

Control: $a = \pi(p, v)$

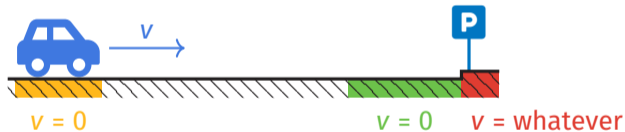
System's Dynamics: $\dot{p} = v$
 $\dot{v} = a$



Deterministic Example

Control: $a = \pi(p, v)$

System's Dynamics: $\dot{p} = v$
 $\dot{v} = a$



General formulation: $u = \pi(x)$

$\dot{x} = f(x, u)$

Deterministic Certificates

How do we certify safety and stability for a deterministic dynamical system?

$$u = \pi(x) \quad \text{and} \quad \dot{x} = f(x, u) \quad \rightarrow \quad \dot{x} = f(x, \pi(x)) = f_{\pi}(x)$$

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Stability

Lyapunov functions V are stability certificates; they satisfy:

- $V(0) = 0$;
- $V(x) > 0$ for $x \neq 0$;
- $\nabla V \cdot f_{\pi}(x) < 0$ for $x \neq 0$.

They certify that the system converges to the equilibrium $x = 0$.

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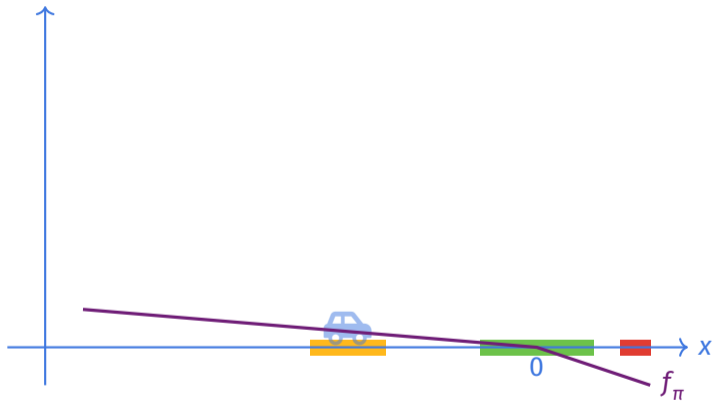
Safety

Control barrier functions V are safety certificates; they satisfy:

- $V(x) \leq \alpha$ for $x \in X_0$;
- $V(x) \geq \beta$ for $x \in X_{\emptyset}$, where $\alpha < \beta$;
- $\nabla V \cdot f_{\pi}(x) \leq -\epsilon$ for $x \in X \setminus X_{\star}$ s.t. $V(x) \leq \beta$.

They certify that the system reaches X_{\star} from X_0 without entering X_{\emptyset} .

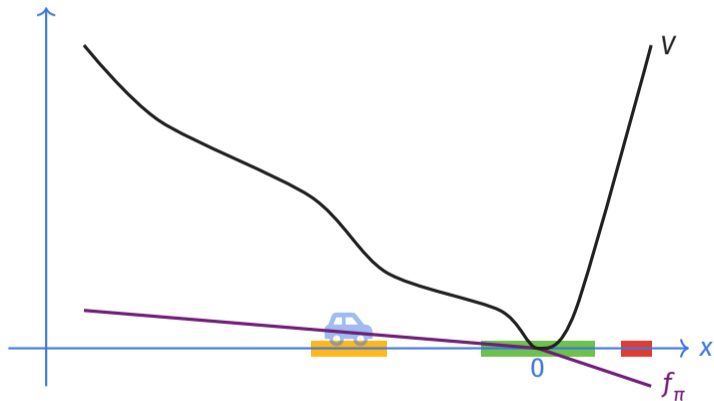
Intuition Behind the Certificates



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Stability

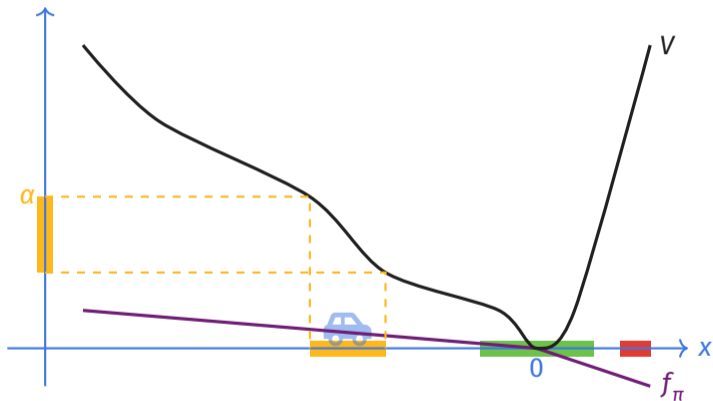
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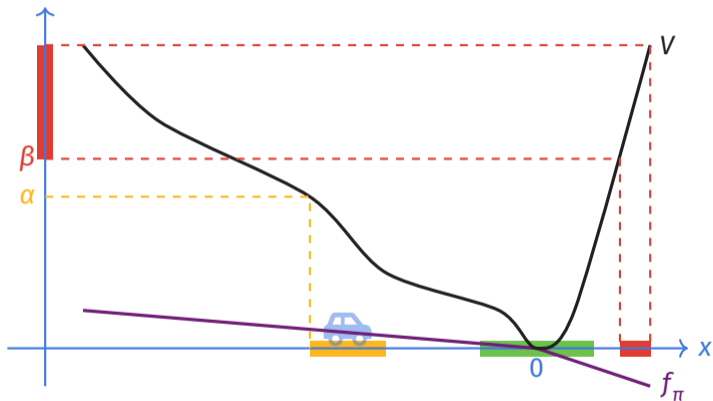
- $V(x) \leq \alpha, x \in \mathbb{X}_0$



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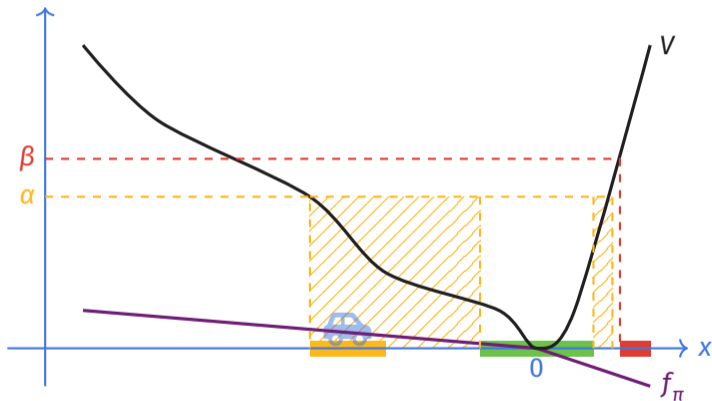
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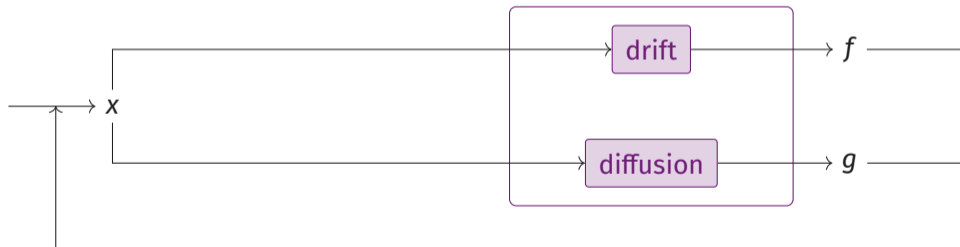
- $V(x) \leq \alpha, x \in \mathbb{X}_0$
- $V(x) \geq \beta, x \in \mathbb{X}_\emptyset$
- $\nabla V \cdot f_\pi(x) \leq -\epsilon,$
 $x \in \mathbb{X} \setminus \mathbb{X}_*$ and
 $V(x) \leq \alpha$



Stochastic Differential Equation

$$dX_t = \underbrace{f(t, X_t)}_{\text{drift}} dt + \underbrace{g(t, X_t)}_{\text{diffusion}} dW_t$$

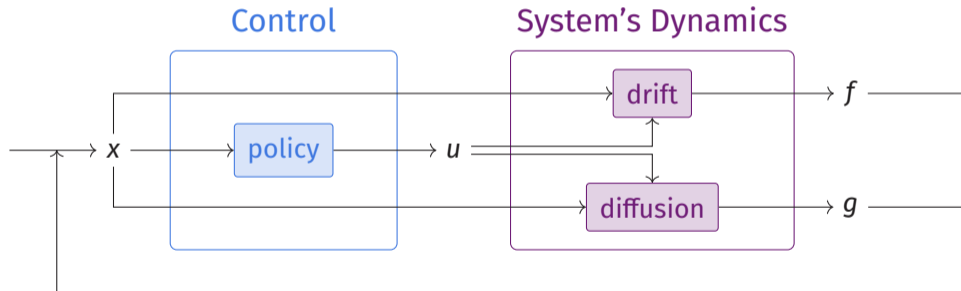
System's Dynamics



Neural Stochastic Differential Equation

$$U_t = \underbrace{\pi(X_t)}_{\text{policy}}$$

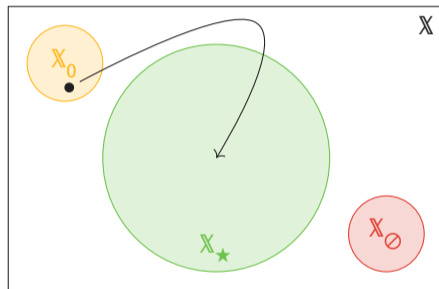
$$dX_t = \underbrace{f(t, X_t, U_t)}_{\text{drift}} dt + \underbrace{g(t, X_t, U_t)}_{\text{diffusion}} dW_t$$



Reach-Avoid Specification

For any initial state $x_0 \in X_0$, certify:

- reach-avoidance with probability at least ϵ .



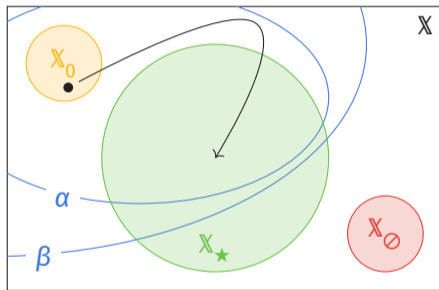
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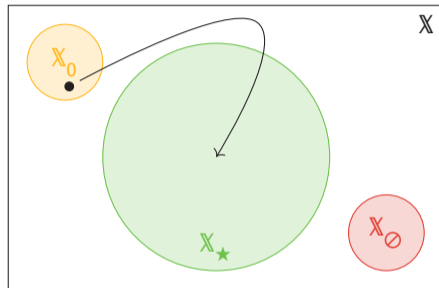
We prove that we can find a certificate and set barriers so that

$$\varepsilon = 1 - \frac{\alpha}{\beta}$$



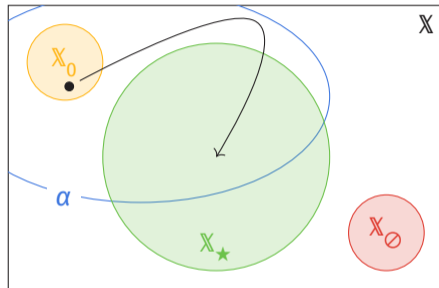
Reach-Avoid Certificate

1. $V(t, x) \geq 0$ for $x \in \mathbb{X}$;



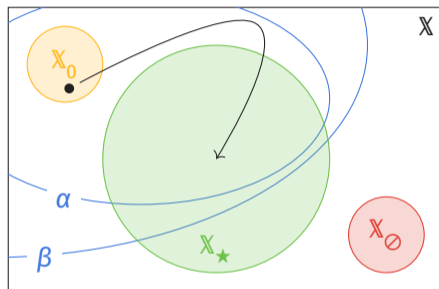
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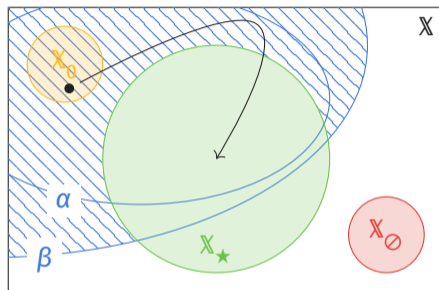
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3. $V(t, x) \geq \beta$ for $x \in \mathbb{X}_\emptyset$;



Reach-Avoid Certificate

1. $V(t, x) \geq 0$ for $x \in \mathbb{X}$;
2. $V(t, x_0) \leq \alpha$ for $x_0 \in \mathbb{X}_0$;
3. $V(t, x) \geq \beta$ for $x \in \mathbb{X}_\emptyset$;
4. $G_\pi V(t, x) < 0$ for $x \notin \text{int } \mathbb{X}_\star$ such that $V(t, x) \leq \beta$.



G_π is the infinitesimal generator (\approx stochastic gradient). It shows the expected change of a function along the state trajectory process.

How Do We Find This Certificate?

We train it using a neural network using the following loss:

$$L = L_{\emptyset} + L_0 + L_{\star} + L_{\downarrow} + R$$

$$L_{\emptyset} = \sum_{x \in B_n \mathbf{x}_{\emptyset}} (\beta - V_{\theta}(x))^+$$

$$L_0 = \sum_{x_0 \in B_n \mathbf{x}_0} (V_{\theta}(x_0) - \alpha)^+$$

$$L_{\star} = \sum_{x \in B_n \mathbf{x}_{\star}} (\beta_S - V_{\theta}(x))^+$$

$$L_{\downarrow} = \sum_{x \in B_n(L_{\beta}^- \setminus L_{\alpha_S}^-)} (G_{\pi} V_{\theta}(x) + \zeta)^+$$

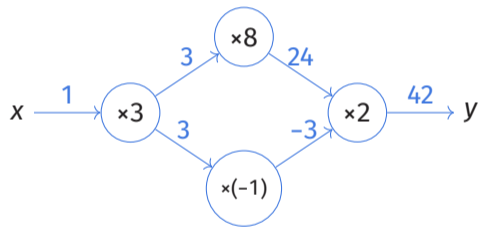
$$R = \lambda \prod_{i=0}^{N-1} \|\mathbf{w}^{(i)}\|_{\infty \rightarrow \infty}$$

The Certificate's Correctness

We use interval bound propagation to verify that the certificate is correct.

The Certificate's Correctness

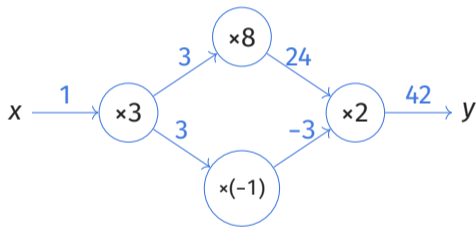
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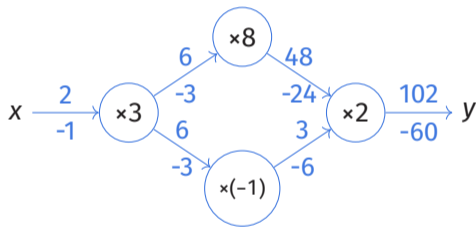
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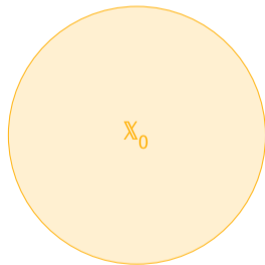


The bounded certificate network

Cell-Splitting for Verification

Verify that $V \leq \alpha$ for all $x_0 \in X_0$

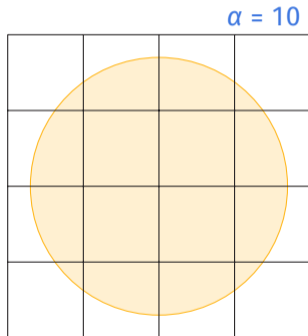
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2. Find the lower and upper bound on V at each cell.

$\alpha = 10$

15	9	9	15
8	7	7	8
9	8	8	9
7	2	2	7
9	8	8	9
7	2	2	7
15	9	9	15
11	7	7	8

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1. Cover X_0 with cells (axis-aligned bounding boxes).
2. Find the **lower** and **upper** bound on V at each cell.
3. A cell is **falsified** if the **lower** bound is above the **threshold**. The considered property does not hold.

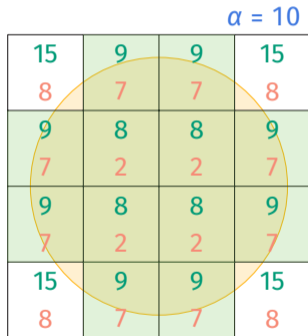
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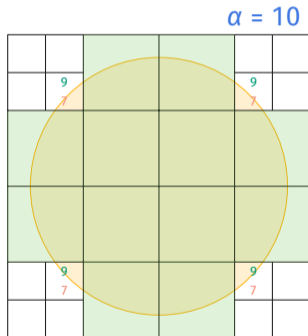
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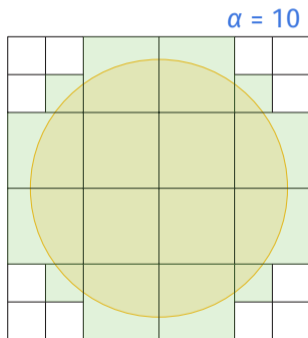
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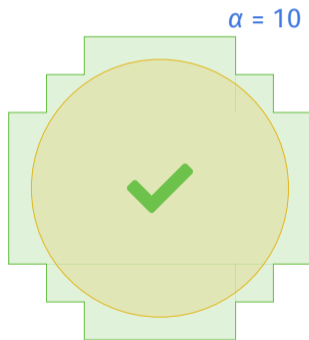
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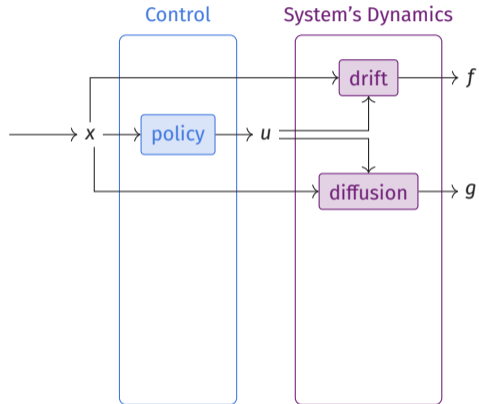
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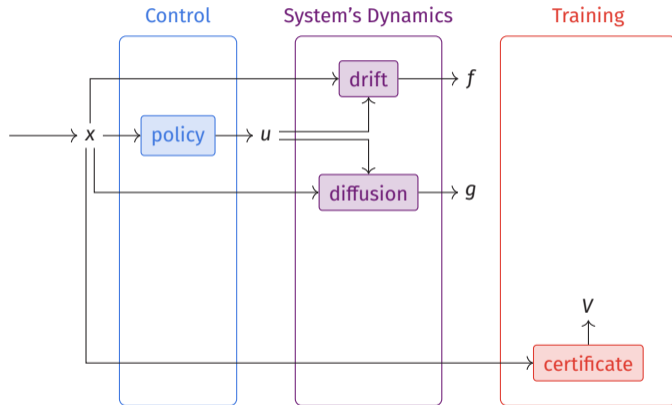
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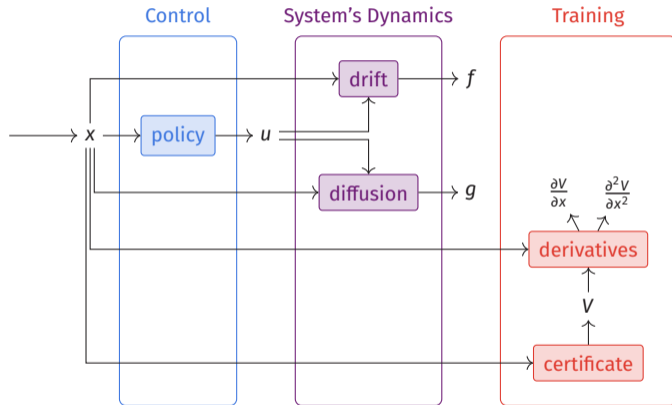
Everything Together All at Once



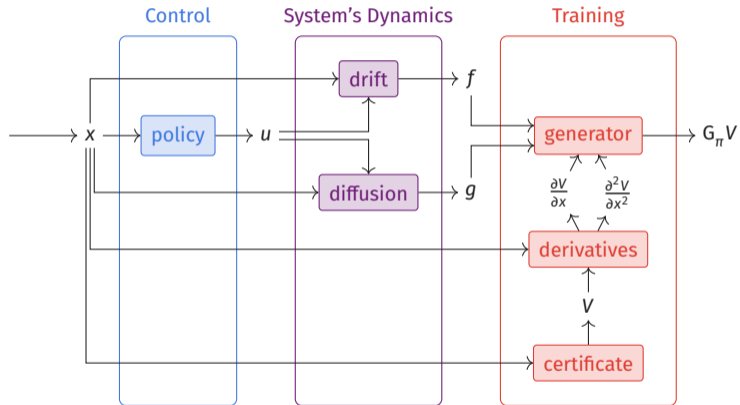
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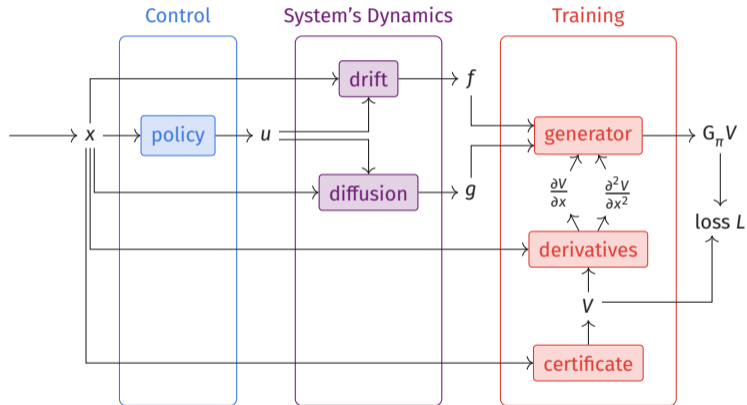
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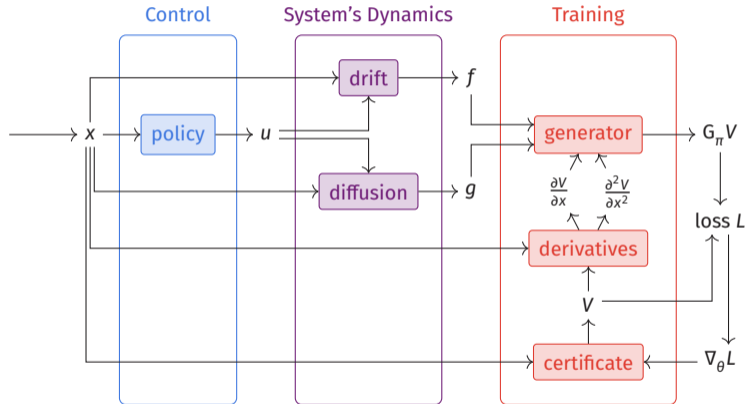
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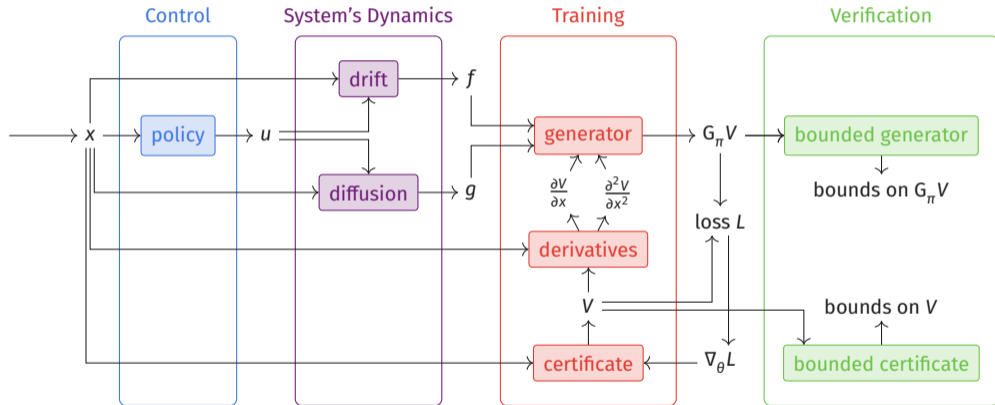
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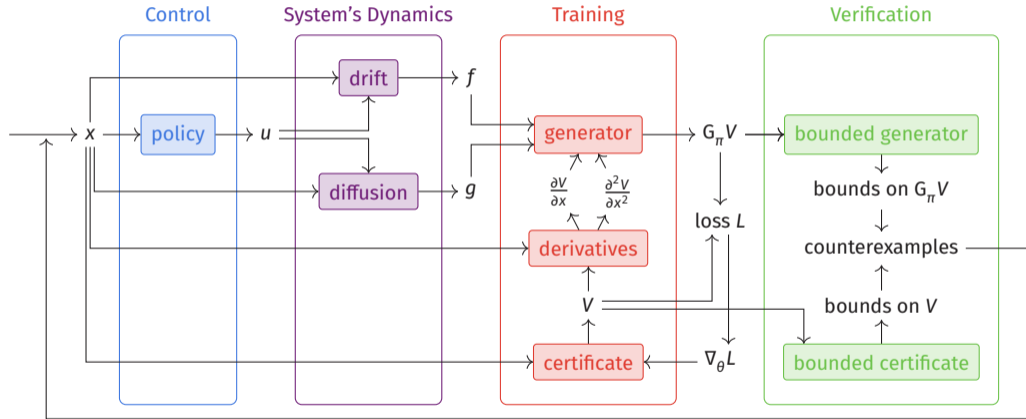
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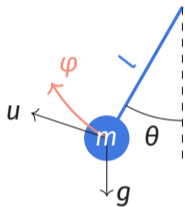
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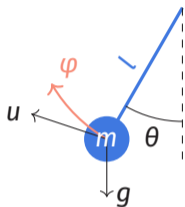


Example — Inverted Pendulum



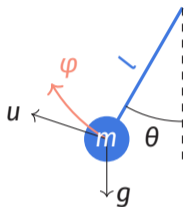
$$\ddot{\theta} = \frac{g}{l} \sin \theta + \frac{Mu - b\dot{\theta}}{ml^2}$$

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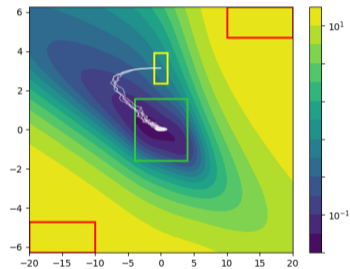


$$d\varphi_t = \left(\frac{g}{l} \sin \theta_t + \frac{Mu_t - b\varphi_t}{ml^2} \right) dt + \sigma dW_t$$
$$d\theta_t = \varphi_t dt$$

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Thank you for your attention!