On ∞ -Safety of Stochastic Differential Dynamics

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Stochasticity

"While writing my book [Stochastic Processes] I had an argument with Feller. He asserted that everyone said 'random variable' and I asserted that everyone said 'chance variable'. We obviously had to use the same name in our books, so we decided the issue by a stochastic procedure. That is, we tossed for it and he won."

[Joseph L. Doob, 1910 - 2004]



Stochasticity in Differential Dynamics



Louis Bachelier

©[Wikipedia]

Brownian motion



Louis Bachelier

@[Wikipedia] Brownian motion

"The mathematical expectation of the speculator is zero."

[L. Bachelier, Théorie de la spéculation, 1900]





Stochasticity in Differential Dynamics



@[Wikipedia] A. Einstein



@[Wikipedia] M. Smoluchowski



P. Langevin



K. Itô



@[Alchetron]

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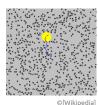
Applications of Stochastic Differential Dynamics



Wind forces



Pedestrian motion



Brownian motion



Population dynamics



Stock options



Robust control

Stochastic Differential Equations (SDEs)

$$dX_t = b(X_t) dt + \sigma(X_t) dW_t, \quad t \ge 0.$$

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Stochastic Differential Equations (SDEs)

Reducing ∞-Safety to T-Safety

$$dX_t = b(X_t) dt + \sigma(X_t) dW_t, \quad t \ge 0.$$

The unique solution is the *stochastic process* $X_t(\omega) = X(t,\omega) : [0,\infty) \times \Omega \to \mathbb{R}^n$ s.t.

$$\label{eq:continuous_def} \textit{X}_{\textit{t}} = \textit{X}_{0} + \int_{0}^{t} \textit{b}(\textit{X}_{\textit{s}}) \; \mathrm{d} \textit{s} + \int_{0}^{t} \sigma(\textit{X}_{\textit{s}}) \; \mathrm{d} \textit{W}_{\textit{s}}.$$

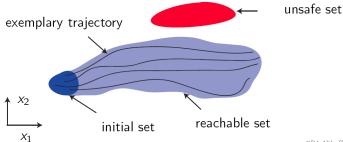
The solution $\{X_t\}$ is also referred to as an (Itô) diffusion process.

Safety Verification of ODEs

Reducing ∞-Safety to T-Safety

Given $T \in \mathbb{R}$, $\mathcal{X} \subseteq \mathbb{R}^n$, $\mathcal{X}_0 \subset \mathcal{X}$, $\mathcal{X}_u \subset \mathcal{X}$, weather

$$\forall \mathbf{x}_0 \in \mathcal{X}_0 \colon \left(\bigcup_{t \leq T} \mathbf{x}_{t, \mathbf{x}_0}\right) \cap \mathcal{X}_u = \emptyset$$
 ?



@[M. Althoff, 2010]

System is T-safe, if no trajectory enters \mathcal{X}_u over [0, T]; Unbounded: $T = \infty$.



Bound the failure probability

$$P\left(\exists t \in [0,\infty) \colon \tilde{\textit{X}}_t \in \mathcal{X}_u
ight), \quad orall \textit{X}_0 \in \{\textit{X} \mid \mathsf{supp}(\textit{X}) \subseteq \mathcal{X}_0\},$$

Experimental Results

Experimental Results

Reducing ∞-Safety to T-Safety

Bound the failure probability

$$P\left(\exists t \in [0,\infty) \colon \tilde{X}_t \in \mathcal{X}_u\right), \quad \forall X_0 \in \{X \mid \mathsf{supp}(X) \subseteq \mathcal{X}_0\},$$

where \tilde{X}_t is the process that will stop at the boundary of \mathcal{X} :

$$ilde{X}_t \, \hat{=} \, extstyle X_{t \wedge au_{\mathcal{X}}} \, = egin{cases} extstyle extstyle X(t,\omega) & ext{if } t \leq au(\omega), \ extstyle X(au(\omega),\omega) & ext{otherwise}, \end{cases}$$

with $\tau_{\mathcal{X}} \cong \inf\{t \mid X_t \notin \mathcal{X}\}.$

∞ -Safety of SDEs

Reducing ∞-Safety to T-Safety

Bound the failure probability

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$$\phi \ \widehat{=} \ " ilde{X}_t$$
 evolves within \mathcal{X} ", $\psi \ \widehat{=} \ " ilde{X}_t$ evolves into \mathcal{X}_u " \Downarrow

 ∞ -safety asks for a bound on $P(\phi \mathcal{U}\psi)$.



Overview of the Idea

Observe that for any $0 \le T < \infty$,

$$P(\exists t \geq 0 \colon \tilde{X}_t \in \mathcal{X}_u) \leq P(\exists t \in [0, T] \colon \tilde{X}_t \in \mathcal{X}_u) + P(\exists t \geq T \colon \tilde{X}_t \in \mathcal{X}_u).$$



Overview of the Idea

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Bounded by an exponential barrier certificate



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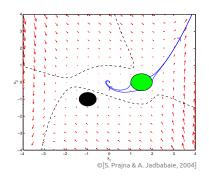
$$P(\exists t \geq 0 \colon \tilde{X}_t \in \mathcal{X}_u) \leq \underbrace{P(\exists t \in [0, T] \colon \tilde{X}_t \in \mathcal{X}_u)}_{\text{Bounded by an exponential barrier certificate}} + \underbrace{P(\exists t \geq T \colon \tilde{X}_t \in \mathcal{X}_u)}_{\text{Bounded by an exponential barrier certificate}}.$$

Bounded by a time-dependent barrier certificate



Recap : Barrier Certificate Witnesses ∞ -Safety

$$\begin{split} & \boldsymbol{\mathcal{B}}(\mathbf{x}) > 0 \quad \forall \mathbf{x} \in \mathcal{X}_{u}, \\ & \boldsymbol{\mathcal{B}}(\mathbf{x}) \leq 0 \quad \forall \mathbf{x} \in \mathcal{X}_{0}, \\ & \frac{\partial \boldsymbol{\mathcal{B}}}{\partial \mathbf{x}}(\mathbf{x}) \boldsymbol{\mathcal{b}}(\mathbf{x}) < 0 \quad \forall \mathbf{x} \in \partial \boldsymbol{\mathcal{B}}. \end{split}$$



Outline

- **1** Reducing ∞ -Safety to *T*-Safety
- 2 Synthesizing Stochastic BCs
- 3 Experimental Results
- **Concluding Remarks**



Bounding the Tail Failure Probability

Definition (Infinitesimal generator [Øksendal, 2013])

Let $\{X_t\}$ be a diffusion process in \mathbb{R}^n . The *infinitesimal generator* \mathcal{A} of X_t is defined by

$$\mathcal{A}f(s,\mathbf{x}) = \lim_{t \downarrow 0} \frac{E^{s,\mathbf{x}} \left[f(s+t,X_t) \right] - f(s,\mathbf{x})}{t}, \quad \mathbf{x} \in \mathbb{R}^n.$$

Let $\mathcal{D}_{\mathcal{A}}$ denote the set of functions for which the limit exists for all $(s, \mathbf{x}) \in \mathbb{R} \times \mathbb{R}^n$.

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Let $\mathcal{D}_{\mathcal{A}}$ denote the set of functions for which the limit exists for all $(s, \mathbf{x}) \in \mathbb{R} \times \mathbb{R}^n$.

Lemma ([Øksendal, 2013])

Let $\{X_t\}$ be a diffusion process defined by an SDE. If $f \in C^{1,2}(\mathbb{R} \times \mathbb{R}^n)$ with compact support, then $f \in \mathcal{D}_{\mathcal{A}}$ and

$$\mathcal{A}f(t,\mathbf{x}) = \frac{\partial f}{\partial t} + \sum_{i=1}^{n} b_i(\mathbf{x}) \frac{\partial f}{\partial x_i} + \frac{1}{2} \sum_{i,j} (\sigma \sigma^{\mathsf{T}})_{ij} \frac{\partial^2 f}{\partial x_i \partial x_j}.$$

 $\mathcal{A}f(t,\mathbf{x})$ generalizes the *Lie derivative* that captures the evolution of $f(t,\mathbf{x})$ along X_t .

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Exponential Stochastic Barrier Certificate

Theorem

Suppose there exists an essentially non-negative matrix $\Lambda \in \mathbb{R}^{m \times m}$, together with an m-dimensional polynomial function (termed exponential stochastic barrier certificate) $V(\mathbf{x}) = (V_1(\mathbf{x}), V_2(\mathbf{x}), \dots, V_m(\mathbf{x}))^T$, with $V_i : \mathbb{R}^n \to \mathbb{R}$ for 1 < i < m, satisfying

$$V(\mathbf{x}) \ge 0 \quad \text{for } \mathbf{x} \in \mathcal{X},$$
 (1)

$$AV(\mathbf{x}) \le -\Lambda V(\mathbf{x}) \quad \text{for } \mathbf{x} \in \mathcal{X},$$
 (2)

$$\Lambda V(\mathbf{x}) \leq \mathbf{0} \quad \text{for } \mathbf{x} \in \partial \mathcal{X}.$$
 (3)

Define a function

$$F(t, \mathbf{x}) \stackrel{\triangle}{=} e^{\Lambda t} V(\mathbf{x}),$$

then every component of $F(t, \tilde{X}_t)$ is a supermartingale.

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Theorem

Reducing ∞-Safety to T-Safety

Bounding the Tail Failure Probability

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Proof

Based on Dynkin's formula [Dynkin, 1965] and Fatou's lemma.



Doob's Supermartingale Inequality

Lemma (Doob's supermartingale inequality [Karatzas and Shreve, 2014])

Let $\{X_t\}_{t>0}$ be a right continuous non-negative supermartingale adapted to a filtration $\{\mathcal{F}_t \mid t>0\}$. Then for any $\lambda>0$,

$$\lambda P\left(\sup_{t\geq 0}X_t\geq \lambda\right)\leq E[X_0].$$

A bound on the probability that a non-negative supermartingale exceeds some given value over a given time interval.



Reducing ∞-Safety to T-Safety

Exponentially Decreasing Bound on the Tail Failure Probability

For cases where $V(\mathbf{x})$ is a scalar function ¹:

Proposition

Suppose there exists a positive constant $\Lambda \in \mathbb{R}$ and a scalar exponential stochastic barrier certificate $V: \mathbb{R}^n \to \mathbb{R}$. Then,

$$P\left(\sup_{t\geq T}V\left(\tilde{X}_{t}\right)\geq \gamma\right)\leq \frac{E\left[V(X_{0})\right]}{\mathrm{e}^{\Lambda T}\gamma}$$

holds for any $\gamma > 0$ and T > 0. Moreover, if there exists l > 0 s.t.

$$V(\mathbf{x}) \geq l$$
 for all $\mathbf{x} \in \mathcal{X}_u$,

then

$$P\left(\exists t \geq T: \tilde{X}_t \in \mathcal{X}_u\right) \leq \frac{E[V(X_0)]}{e^{\Lambda T_l}}$$

holds for any T > 0.



Experimental Results

Exponentially Decreasing Bound on the Tail Failure Probability

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holds for any T > 0.

Proof

Based on Doob's supermartingale inequality.

1. The result generalizes to the slightly more involved case where $V(\mathbf{x})$ is a vector function.

Exponentially Decreasing Bound on the Tail Failure Probability

 $\forall \epsilon > 0. \,\exists \tilde{\tau} > 0:$ the truncated $\tilde{\tau}$ -tail failure probability is bounded by ϵ :

Theorem

If there exists $\alpha > 0$, s.t. $\forall \mathbf{x} \in \mathcal{X}_0 : V_i(\mathbf{x}) < \alpha$ holds for some $i \in \{1, \dots, m\}$. Then for any $\epsilon > 0$, there exists $\tilde{T} > 0$ s.t.

$$P\left(\exists t \geq \tilde{T}: \tilde{X}_t \in \mathcal{X}_u\right) \leq \epsilon.$$

Theorem

Bounding the Failure Probability over [0, T]

Suppose there exists a constant $\eta > 0$ and a polynomial function (termed time-dependent stochastic barrier certificate) $H(t, \mathbf{x}) : \mathbb{R} \times \mathbb{R}^n \to \mathbb{R}$, satisfying

$$H(t, \mathbf{x}) > 0$$
 for $(t, \mathbf{x}) \in [0, T] \times \mathcal{X}$, (4)

$$\mathcal{A}H(t,\mathbf{x}) \le 0 \quad \text{for } (t,\mathbf{x}) \in [0,T] \times (\mathcal{X} \setminus \mathcal{X}_u),$$
 (5)

$$\frac{\partial H}{\partial t} \le 0 \quad for(t, \mathbf{x}) \in [0, T] \times \partial \mathcal{X},$$
 (6)

$$H(t, \mathbf{x}) \ge \eta \quad \text{for } (t, \mathbf{x}) \in [0, T] \times \mathcal{X}_u.$$
 (7)

Then.

$$P\left(\exists t \in [0,T] \colon \tilde{X}_t \in \mathcal{X}_u\right) \leq \frac{E[H(0,X_0)]}{n}.$$

Time-Dependent Stochastic Barrier Certificate

Theorem

Suppose there exists a constant $\eta > 0$ and a polynomial function (termed time-dependent stochastic barrier certificate) $H(t, \mathbf{x}) : \mathbb{R} \times \mathbb{R}^n \to \mathbb{R}$, satisfying

$$H(t, \mathbf{x}) \ge 0 \quad \text{for } (t, \mathbf{x}) \in [0, T] \times \mathcal{X},$$
 (4)

$$\mathcal{A}H(t,\mathbf{x}) \le 0 \quad \text{for } (t,\mathbf{x}) \in [0,T] \times (\mathcal{X} \setminus \mathcal{X}_u),$$
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$$\frac{\partial H}{\partial t} \le 0 \quad \text{for } (t, \mathbf{x}) \in [0, T] \times \partial \mathcal{X},$$
 (6)

$$H(t, \mathbf{x}) \ge \eta \quad \text{for } (t, \mathbf{x}) \in [0, T] \times \mathcal{X}_u.$$
 (7)

Then.

$$P\left(\exists t \in [0,T] \colon \tilde{X}_t \in \mathcal{X}_u\right) \leq \frac{E[H(0,X_0)]}{\eta}.$$

Proof

Based on Dynkin's formula and Doob's supermartingale inequality.



Time-Dependent Stochastic Barrier Certificate

Corollary

Suppose there exists $\beta > 0$, s.t. $H(0, \mathbf{x}) \leq \beta$ for $\mathbf{x} \in \mathcal{X}_0$. Then,

$$P(\exists t \in [0, T]: \tilde{X}_t \in \mathcal{X}_u) \leq \frac{\beta}{\eta}.$$

SDP Encoding for Synthesizing V(x)

Reducing ∞-Safety to T-Safety

minimize
$$\alpha$$
 (8)

subject to
$$V^a(\mathbf{x}) \geq 0$$
 for $\mathbf{x} \in \mathcal{X}$ (9)

$$\mathcal{A}V^{a}(\mathbf{x}) \leq -\Lambda V^{a}(\mathbf{x}) \quad \text{for } \mathbf{x} \in \mathcal{X}$$
 (10)

$$\Lambda V^{a}(\mathbf{x}) \le 0 \quad \text{for } \mathbf{x} \in \partial \mathcal{X} \tag{11}$$

$$V^a(\mathbf{x}) \ge 1 \quad \text{for } \mathbf{x} \in \mathcal{X}_u$$
 (12)

$$V^{a}(\mathbf{x}) < \alpha \mathbf{1} \quad \text{for } \mathbf{x} \in \mathcal{X}_{0}$$
 (13)



SDP Encoding for Synthesizing $H(t, \mathbf{x})$

Reducing ∞-Safety to T-Safety

$$\min_{b,\beta} \text{minimize } \beta \tag{14}$$

subject to
$$H^b(t, \mathbf{x}) \ge 0$$
 for $(t, \mathbf{x}) \in [0, 7] \times \mathcal{X}$ (15)

$$\mathcal{A}H^{b}(t,\mathbf{x}) \leq 0 \quad \text{for } (t,\mathbf{x}) \in [0,T] \times (\mathcal{X} \setminus \mathcal{X}_{u})$$
 (16)

$$\frac{\partial \mathcal{H}^b}{\partial t} \le 0 \quad \text{for } (t, \mathbf{x}) \in [0, T] \times \partial \mathcal{X}$$
 (17)

$$H^b(t, \mathbf{x}) \ge 1$$
 for $(t, \mathbf{x}) \in [0, T] \times \mathcal{X}_u$ (18)

$$H^{b}(0,\mathbf{x}) \le \beta$$
 for $\mathbf{x} \in \mathcal{X}_{0}$ (19)



•0

Example: Population Dynamics

Example (Population growth [Panik, 2017])

$$dX_t = -X_t dt + \sqrt{2}/2X_t dW_t.$$

$$\infty\text{-safety setting:}\ \mathcal{X}=\{\mathbf{x}\mid\mathbf{x}\geq0\}, \mathcal{X}_0=\{\mathbf{x}\mid\mathbf{x}=1\}, \mathcal{X}_{\textit{u}}=\{\mathbf{x}\mid\mathbf{x}\geq2\}.$$

Example: Population Dynamics

Reducing ∞-Safety to T-Safety

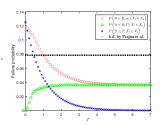
Example (Population growth [Panik, 2017])

$$\mathrm{d}X_t = -X_t \; \mathrm{d}t + \sqrt{2}/2X_t \; \mathrm{d}W_t.$$

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$$\begin{split} \textit{V}(\mathbf{x}) &= 0.000001630047868 - 0.000048762786972\mathbf{x} \\ &+ 0.125025533525219\mathbf{x}^2 + 0.000000001603294\mathbf{x}^3. \end{split}$$

$$P\left(\exists t \geq T \colon \tilde{X}_t \in \mathcal{X}_u\right) \leq \frac{0.12498}{\mathrm{e}^T} \quad \forall T > 0.$$



Example: Harmonic Oscillator

Reducing ∞-Safety to T-Safety

Example (Harmonic oscillator [Hafstein et al., 2018])

$$dX_t = \begin{pmatrix} 0 & \omega \\ -\omega & -k \end{pmatrix} X_t dt + \begin{pmatrix} 0 & 0 \\ 0 & -\sigma \end{pmatrix} X_t dW_t.$$

Constants : $\omega = 1$, k = 7, $\sigma = 2$. ∞-safety setting : $\mathcal{X} = \mathbb{R}^n$, $\mathcal{X}_0 = \{(\mathbf{x}_1, \mathbf{x}_2) \mid -1.2 \le \mathbf{x}_1 \le 0.8, -0.6 \le \mathbf{x}_2 \le 0.4\}$, $\mathcal{X}_u = \{(\mathbf{x}_1, \mathbf{x}_2) \mid |\mathbf{x}_1| \ge 2\}$.

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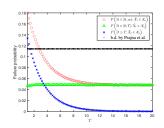
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$$\forall T > 1$$
:

Reducing ∞-Safety to T-Safety

$$P\left(\exists t \geq T : \tilde{X}_t \in \mathcal{X}_u\right) \leq \frac{0.19927}{0.00005e^{0.2T} + 1.00025e^{0.4T}}.$$





Summary

Reducing ∞-Safety to T-Safety

For any $0 \leq T < \infty$,

$$P(\exists t \geq 0 \colon \tilde{X}_t \in \mathcal{X}_u) \leq \underbrace{P(\exists t \in [0, T] \colon \tilde{X}_t \in \mathcal{X}_u)}_{\text{Bounded by an exponential barrier certificate}} + \underbrace{P(\exists t \geq T \colon \tilde{X}_t \in \mathcal{X}_u)}_{\text{Bounded by an exponential barrier certificate}}.$$

Bounded by a time-dependent barrier certificate



SDEs with control inputs? ∞ -Safety of Probabilistic Programs?

