

Peak Value-At-Risk Estimation for Stochastic Differential Equations Using Occupation Measures

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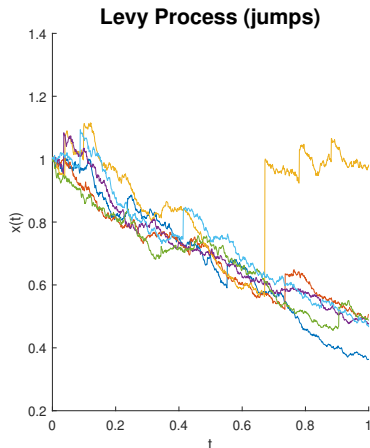
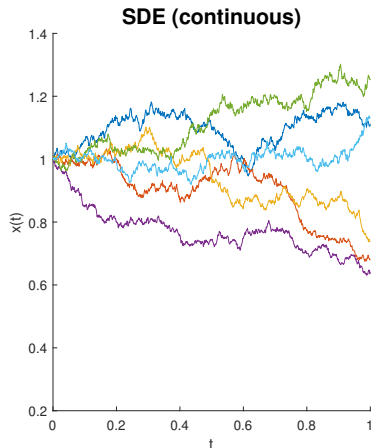
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ThB22.6: IEEE CDC



Stochastic Process

Nonanticipative, indep.-increment set of prob. dists. $\{\mu_t\}$



Geometric Brownian motion (left), Merton jump diffusion (right)

Questions to ask

Given a state function $p(x)$ (e.g. height, voltage):

What is the maximum along stochastic trajectories of the:

- Mean of p ?
- Quantile statistics of p ?
- Conditional Value-at-Risk of p ?

Risk, safety analysis.

Examples of Generators

Generator \mathcal{L} of process $\forall v \in \text{dom}(\mathcal{L}) = \mathcal{C}$:

$$\mathcal{L}_\tau v = \lim_{\tau' \rightarrow \tau} (\mathbb{E}[v(t + \tau', x) \mid \mu_{t+\tau'}] - v(t, x)) / \tau' \quad (1)$$

Discrete-time Markov Process ($\mathcal{C} = C([0, T] \times X)$)

$$X_{t+\tau} = F(t, X_t, \omega_t), \quad \omega_t \sim \xi \text{ (sampled)} \quad (2)$$

$$\mathcal{L}_\tau v = \left(\int_{\Omega} v(t + \tau, F(t, x, \omega)) d\xi(\omega) - v \right) / \tau \quad (3)$$

Examples of Generators (Continuous-time)

Generator \mathcal{L} of process $\forall v \in \text{dom}(\mathcal{L}) = \mathcal{C}$:

$$\mathcal{L}_\tau v = \lim_{\tau' \rightarrow \tau} (\mathbb{E}[v(t + \tau', x) \mid \mu_{t+\tau'}] - v(t, x)) / \tau' \quad (4)$$

Stochastic Differential Equation ($\mathcal{C} = C^{1,2}([0, T] \times X)$)

$$dx = f(t, x)dt + g(t, x)dW, \quad (5)$$

$$\mathcal{L}_0 v = \partial_t v + f \cdot \nabla_x v + g^T (\nabla_{xx}^2 v) g / 2 \quad (6)$$

Others: Lévy processes, hybrid, switching, time-delay

Chance-Peak Problem

Distribution $p_{\#}\mu_t$ of $p(x(t))$ (pushforward)

What is the maximum risk R along the stochastic trajectory?

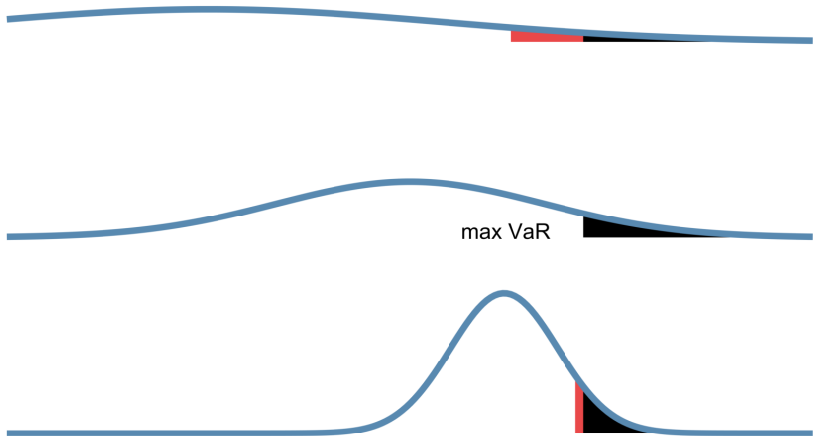
$$P^* = \sup_{t^* \in [0, T]} R(p_{\#}\mu_{t^*}) \quad (7a)$$

$$x(t) \text{ follows } \mathcal{L} \quad \forall t \in [0, t^*] \quad (7b)$$

$$x(0) \sim \mu_0 \quad (7c)$$

Maximal Value at Risk

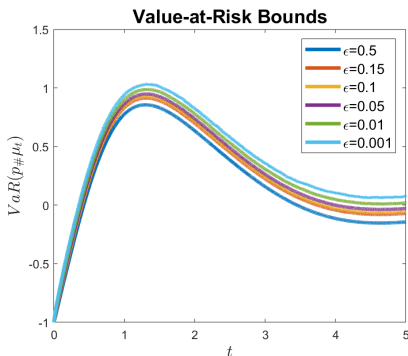
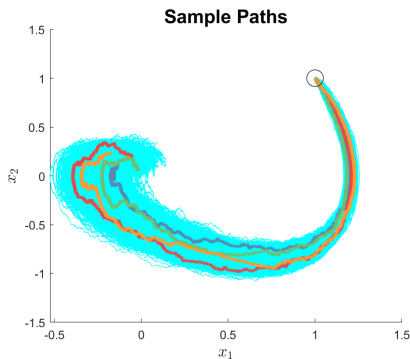
Maximize ϵ -VaR among multiple distributions



Red + Black areas = 10% probability

Value-at-Risk Example (Monte Carlo)

50,000 samples with $T = 5$, $\Delta t = 10^{-3}$



$$\text{VaR of } p = -x_2 \text{ along } dx = \begin{bmatrix} x_2 \\ -x_1 - x_2 - \frac{1}{2}x_1^3 \end{bmatrix} dt + \begin{bmatrix} 0 \\ 0.1 \end{bmatrix} dw$$

Chance-Peak Measure Programs

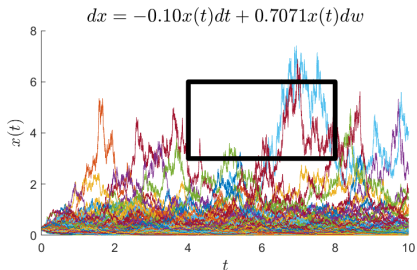
Occupation measures

Avg. time trajectories spend in set

Test function $v(t, x) \in \mathcal{C}$

Averaged value:

$$\langle v, \mu \rangle = \int_0^T \mathbb{E}_{x \sim X_t} [v(t, x)] dt$$



Martingale Relation

$$\begin{aligned} \forall v : \mathbb{E}[v(t+s, x) \mid X_{t+s}] &= \mathbb{E}[v(t, x) \mid X_t] \\ &+ \int_{s'=t}^{t+s} \mathbb{E}[\mathcal{L}_0 v(t, x) \mid X_{s'}] \quad (8) \end{aligned}$$

Relation between measures (μ_t, μ_{t+s}, μ)

$$\langle v(t+s, x), \mu_{t+s} \rangle = \langle v(t, x), \mu_t \rangle + \langle \mathcal{L}_0 v(t', x), \mu \rangle \quad (9)$$

Shorthand notation (adjoint)

$$\mu_{t+s} = \mu_t + \mathcal{L}_0^\dagger \mu \quad (10)$$

Triple of (9) is supported on graph of \mathcal{L}_0 (assuming compact + regularity), similar for [sum of \mathcal{L}_τ].

Mean Maximization

Infinite-dimensional Linear Program (Cho, Stockbridge, 2002)

$$p^* = \sup \langle p, \mu_p \rangle \quad (11a)$$

$$\mu_p = \delta_0 \otimes \mu_0 + \mathcal{L}_f^\dagger \mu \quad (11b)$$

$$\mu, \mu_p \in \mathcal{M}_+([0, T] \times X) \quad (11c)$$

Instance of Optimal Control Program (Lewis and Vinter, 1980)

(μ_p^*, μ^*) is feasible with $P^* = \langle p(x), \mu_p^* \rangle \leq p^*$

$P^* = p^*$ if compactness, regularity properties hold

Value-at-Risk Bounds

VaR is nonconvex, nonsubadditive

Concentration inequalities can upper-bound VaR

$$\text{VaR}_\epsilon(\xi) \leq \text{stdev}(\xi)r + \text{mean}(\xi)$$

Name	r	Valid Condition
Cantelli	$\sqrt{1/(\epsilon) - 1}$	ξ probability distribution
VP	$\sqrt{4/(9\epsilon) - 1}$	ξ unimodal, $\epsilon < 1/6$

Conditional Value at Risk (**CVaR**) can also bound VaR

Concentration-Bounded Chance-Peak

Apply concentration inequalities to get upper bound $P_r^* \geq P^*$

Objective upper-bounds VaR w.r.t. time- t^* distribution μ_{t^*}

$$P_r^* = \sup_{t^* \in [0, T]} r \sqrt{\langle p^2, \mu_{t^*} \rangle - \langle p, \mu_{t^*} \rangle^2} + \langle p, \mu_{t^*} \rangle \quad (12a)$$

$$x \text{ follows } \mathcal{L} \quad (12b)$$

$$x(0) \sim \mu_0 \quad (12c)$$

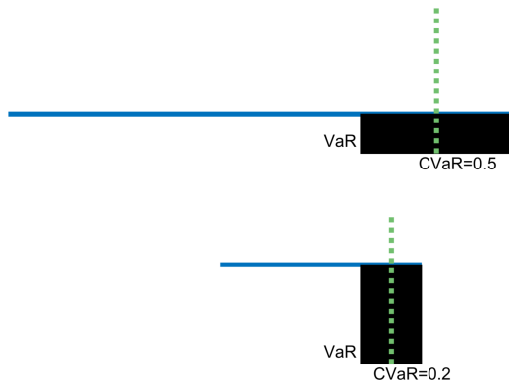
SOCP in measures (3d SOC) for $p_r^* \geq P_r^*$

Same constraints as mean-maximization, different objective

Conditional Value-at-Risk

CVAR: Average quantity above the Value-at-Risk

$$CVaR_{\epsilon}(\xi(\omega)) = (1/\epsilon) \int_{\omega \geq VaR_{\epsilon}(\xi)} \omega d\xi(\omega)$$



Uniform distributions with same VaR, different CVaR (70%)

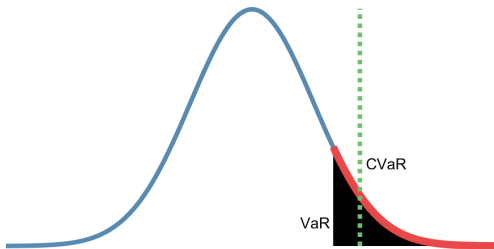
CVaR Linear Program

Measure LP to compute CVaR (with $\frac{d\psi}{d\xi} \leq \frac{1}{\epsilon}$)

$$CVaR_{\epsilon}(\nu) = \sup_{\psi, \hat{\psi} \in \mathcal{M}_+(\mathbb{R})} \text{mean}(\psi) \quad (13a)$$

$$\epsilon\psi + \hat{\psi} = \nu \quad (13b)$$

$$\langle 1, \psi \rangle = 1 \quad (13c)$$



$$\text{VaR} = 1.2816, \text{ CVaR} = 1.7550, \epsilon\psi \leq \xi$$

Highest CVaR along SDE trajectories

$$P_c^* = \sup_{t^* \in [0, T]} \text{CVaR}_\epsilon(p_{\#} \mu_{t^*}) \quad (14a)$$

$$x \text{ follows } \mathcal{L} \quad (14b)$$

$$x(0) \sim \mu_0 \quad (14c)$$

Almost the same as VaR chance-peak, with $P_c^* \geq P^*$

CVaR Measure program

Add CVaR objective, constraints to chance-peak

$$p_c^* = \sup \quad \text{mean}(\psi) \quad (15a)$$

$$\mu_\tau = \delta_0 \otimes \mu_0 + \mathcal{L}^\dagger \mu \quad (15b)$$

$$\langle \mathbf{1}, \psi \rangle = 1 \quad (15c)$$

$$\epsilon \psi + \hat{\psi} = p_{\#} \mu_\tau \quad (15d)$$

$$\mu, \mu_\tau \in \mathcal{M}_+([0, T] \times X) \quad (15e)$$

$$\psi, \hat{\psi} \in \mathcal{M}_+(\mathbb{R}) \quad (15f)$$

Upper-bound $p_c^* \geq P_c^* \geq P^*$, LP in measures

Comparison of bounds

$P_r^* = p_r^*$ and $P_c^* = p_c^*$ if

1. \mathcal{L} has unique solutions (e.g. SDE: Lipchitz, Growth)
2. $[0, T] \times X$ compact
3. $p(x)$ is continuous

$P_{\text{Cantelli}}^* \geq P_c^*$ always, but (P_c^*, P_{VP}^*) incomparable (so far)

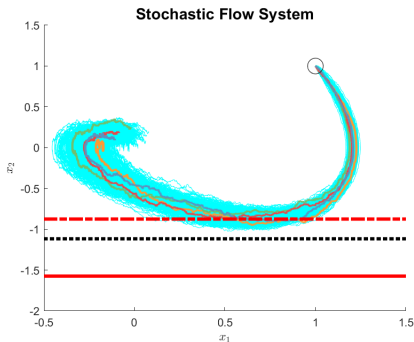
Empirically, degree- k moment LMLs satisfy $p_{\text{Cantelli},k}^* \geq p_{c,k}^*$

Chance-Peak Examples

Two-State

Stochastic Flow (Prajna, Rantzer) with $T = 5$, $p(x) = -x_2$

$$dx = \begin{bmatrix} x_2 \\ -x_1 - x_2 - \frac{1}{2}x_1^3 \end{bmatrix} dt + \begin{bmatrix} 0 \\ 0.1 \end{bmatrix} dw$$

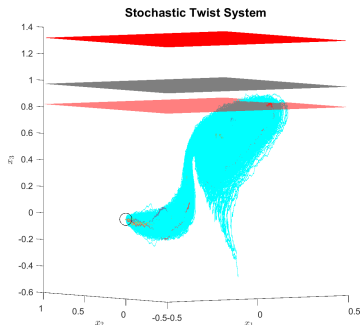


$d = 6$ (dash-dot=50%, dotted-black=85% CVAR, solid=85% VP)

Three-State

Stochastic Twist system with $T = 5$, $p(x) = x_3$

$$dx = \begin{bmatrix} -2.5x_1 + x_2 - 0.5x_3 + 2x_1^3 + 2x_3^3 \\ -x_1 + 1.5x_2 + 0.5x_3 - 2x_2^3 - 2x_3^3 \\ 1.5x_1 + 2.5x_2 - 2x_3 - 2x_1^3 - 2x_2^3 \end{bmatrix} dt + \begin{bmatrix} 0 \\ 0 \\ 0.1 \end{bmatrix} dw$$

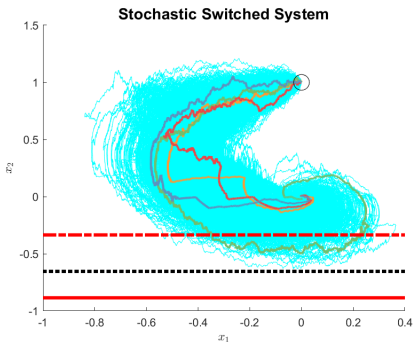


$d = 6$ (translucent=50%, gray=85% CVAR, solid=85% VP)

Two-State Switching

Switching subsystems at $T = 5$, $p(x) = -x_2$

$$dx = \left\{ \begin{bmatrix} -2.5x_1 - 2x_2 \\ -0.5x_1 - x_2 \end{bmatrix}, \begin{bmatrix} -x_1 - 2x_2 \\ 2.5x_1 - x_2 \end{bmatrix} \right\} dt + \begin{bmatrix} 0 \\ 0.25x_2 \end{bmatrix} dw$$

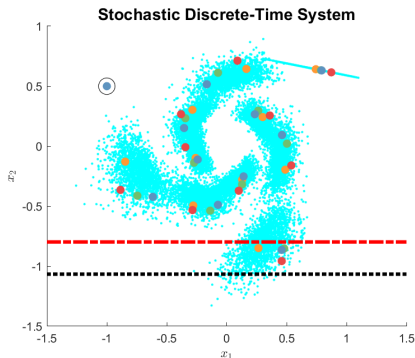


$d = 6$ (dash-dot=50%, dotted-black=85% CVAR, solid=85% VP)

Two-State Discrete-Time

Parameter λ sampled from $\lambda[t] \in \mathcal{N}(0, 1)$

$$x_+ = \begin{bmatrix} -0.3x_1 + 0.8x_2 + x_1x_2\lambda/4 \\ -0.9x_1 - 0.1x_2 - 0.2x_1^2 + \lambda/40 \end{bmatrix}.$$



$d = 6$ (dash-dot=50%, black-dotted=85% CVaR)

Take-aways

Conclusion

Posed the chance-peak problem for wide class of \mathcal{L}

Solved using infinite-dimensional SOCPs, LPs in measures

Certified outer-approximations of risk

Acknowledgements

Roy Smith, Automatic Control Lab (IfA)

POP group at LAAS-CNRS

NCCR Automation

Air Force Office for Scientific Research

National Science Foundation

Thanks!

Questions?

$$d_{\mathbb{E}}^* = \min \int_X v(0, x) \, d\mu_0(x) \quad (16a)$$

$$- \mathcal{L}v(t, x) \in \Sigma[[0, T] \times X] \quad (16b)$$

$$v(t, x) - p(x) \in \Sigma[[0, T] \times X] \quad (16c)$$

$$v \in \mathbb{R}[t, x]. \quad (16d)$$

SOS Concentration-Peak

$$d_r^* = \inf \quad u_1 + 2u_3 + \int_{X_0} v(0, x_0) d\mu_0(x_0) \quad (17a)$$

$$- \mathcal{L}v(t, x) \in \Sigma[[0, T] \times X] \quad (17b)$$

$$v(t, x) + u_1 p^2(x) - 2u_2 p(x) - p(x) \quad (17c)$$

$$\in \Sigma[[0, T] \times X]$$

$$([u_1 + u_3, -(r/2), u_2], u_3) \in \mathbb{L}^3 \quad (17d)$$

$$u \in \mathbb{R}^3, \quad v \in \mathcal{C}([0, T] \times X).$$

$$d_c^* = \min \quad u + \int_X v(0, x) \, d\mu_0(x) \quad (18a)$$

$$- \mathcal{L}v(t, x) \in \Sigma[[0, T] \times X] \quad (18b)$$

$$v(t, x) - w(p(x)) \in \Sigma[[0, T] \times X] \quad (18c)$$

$$u + \epsilon w(q) - q \in \Sigma[p_{min}, p_{max}] \quad (18d)$$

$$w(q) \in \Sigma[p_{min}, p_{max}] \quad (18e)$$

$$u \in \mathbb{R}, v \in \mathbb{R}[t, x]. \quad (18f)$$