

# Stochastic invariance of closed sets with non-Lipschitz coefficients

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# Problem

## Aim of this work

$(b, \sigma) : \mathbb{R}^d \rightarrow \mathbb{R}^d \times \mathbb{M}^d$  continuous,  $b$  and  $\|\sigma\sigma^\top\|^{\frac{1}{2}}$  with linear growth.

$$X = x + \int_0^\cdot b(X_s) ds + \int_0^\cdot \sigma(X_s) dW_s$$

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Necessary and sufficient conditions for the existence of a solution  $X \in \mathcal{D}$ , given  $x \in \mathcal{D}$  a closed set, i.e.  $\mathcal{D}$  is stochastically invariant.

# Litterature

- **General answer for “smooth coefficients”** : Friedman [6], Doss [3], Bardi and Goatin [4] and Bardi and Jensen [5] (2nd order normal cone). Da Prato and Frankowska [1] and Buckdahn et al. [7] (first order normal cone), Tappe [10] (jump diffusions).
- **Affine or polynomial models using specific treatments** : Filipović and Mayerhofer [5], Filipović and Larsson [4] (polynomial diffusions), Cuchiero *et al.* [10] (affine processes on the cone of symmetric semi-definite matrices), Spreij and Veerman [9] (affine diffusions).

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⇒ We want a general answer using the first order normal cone covering smooth and non-smooth coefficients (e.g.  $\sigma(x) = \sqrt{x}$ )

Replace  $\sigma \in C_b^{1,1}$  by  $\sigma\sigma^\top \in C_{\text{loc}}^{1,1}$   
and use the first order normal cone.

The regular case :  $\sigma \in C_{loc}^{1,1}$

Da Prato and Frankowska [1]  
and Buckdahn, Quincampoix, Rainer & Teichmann [7]

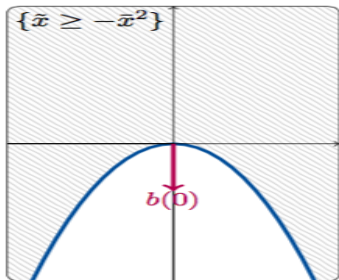


## Characterization in the regular case

**Thm** : Assume that  $\sigma \in C_{loc}^{1,1}$ .  $\mathcal{D}$  is stochastically invariant if and only if

$$\sigma(x)^\top u = 0 \text{ and } \langle u, b(x) - \frac{1}{2} \sum_{j=1}^d D\sigma^j(x)\sigma^j(x) \rangle \leq 0, \forall x \in \mathcal{D} \text{ and } u \in \mathcal{N}_{\mathcal{D}}^1(x),$$

where  $\mathcal{N}_{\mathcal{D}}^1(x) := \{u \in \mathbb{R}^d : \langle u, y - x \rangle \leq o(\|y - x\|), \forall y \in \mathcal{D}\}$  is the first order normal cone at  $x$ .



## Necessary condition in the regular case

Suffices to check that  $\phi(X) \leq 0$  for  $\phi : y \mapsto \langle u, y - x \rangle - \frac{\kappa}{2} \|y - x\|^2$ , for some  $\kappa > 0$ .

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a. Apply Itô's Lemma and Girsanov theorem to get

$$0 \geq \int_0^t [\mathcal{L}\phi(X_s) + n \|D\phi(X_s)\sigma(X_s)\|^2] ds + \int_0^t D\phi(X_s)\sigma(X_s) dW_s^n.$$

Take expectation under  $\mathbb{P}^n$ , divide by  $t$  and  $t \rightarrow 0$  :

$$D\phi(x)\sigma(x) = 0 \Leftrightarrow \sigma(x)^\top u = 0.$$

## Necessary condition in the regular case

b. Apply Itô's Lemma twice :

$$\begin{aligned} 0 &\geq \int_0^t \mathcal{L}\phi(X_s)ds + \int_0^t D\phi(X_s)\sigma(X_s)dW_s \\ &= \int_0^t \mathcal{L}\phi(X_s)ds + \int_0^t \left[ D\phi(x)\sigma(x) + \int_0^s \mathcal{L}(D\phi\sigma)(X_u)du \right] dW_s \\ &\quad + \int_0^t \int_0^s D(D\phi\sigma)(X_u)\sigma(X_u)dW_u dW_s \end{aligned}$$

## Necessary condition.

Use Cheridito, Soner & Touzi [8], Bruder [6], Buckdahn et al. [7] ( $d = 1$  here). Since  $X$  is Hölder continuous and the functions are continuous :

$$\begin{aligned} 0 \geq & \mathcal{L}\phi(x)t + \underbrace{\int_0^t [\mathcal{L}\phi(X_s) - \mathcal{L}\phi(x)] ds}_{o(t)} + \underbrace{\int_0^t \int_0^s \mathcal{L}(D\phi\sigma)(X_u) dudW_s}_{O(t^{\frac{3}{2}-\varepsilon})} \\ & + D(D\phi\sigma)(x)\sigma(x) \frac{W_t^2 - t}{2} \\ & + \underbrace{\int_0^t \int_0^s [D(D\phi\sigma)(X_u)\sigma(X_u) - D(D\phi\sigma)(x)\sigma(x)] dW_u dW_s}_{O(t^{1+\eta})}. \end{aligned}$$

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Divide by  $t$  and use that  $\liminf_{t \rightarrow 0} W_t^2/t = 0$  ( $d = 1$  here) :

$$0 \geq \mathcal{L}\phi(x) - \frac{1}{2} D(D\phi\sigma)(x)\sigma(x) = \langle u, b(x) - \frac{1}{2} D\sigma(x)\sigma(x) \rangle.$$

Exemple of irregular case :  $\sigma(x) = \sqrt{|x|}$

## The case $\sigma(x) = \sqrt{|x|}$ and $X \geq 0$

We consider

$$X = 0 + \int_0^\cdot a(b - X_s) ds + \int_0^\cdot \sqrt{|X_s|} dW_s.$$



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Can not apply Itô's Lemma to  $\sqrt{|X|}$ ...  
but can just take expectation to get

$$0 \leq \lim_{t \rightarrow 0} \frac{1}{t} \mathbb{E} \left[ \int_0^t a(b - X_s)ds \right] = ab,$$

that turns out to be necessary and sufficient....

## Regular vs Irregular case

- Regular case : want to keep the contribution of the diffusion part  $\Rightarrow$  “pathwise analysis”.

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Need to find a way to  
kill the “irregular” directions and keep the “regular” ones.

The general case  
 $\sigma\sigma^\top$  can be extended into a  $C_{loc}^{1,1}$  function  $C$

## A toy example

We consider

$$X = 0 + \int_0^\cdot b(X_s) ds + \int_0^\cdot \underbrace{\begin{pmatrix} \sigma^1(X_s) & 0 \\ 0 & \sqrt{X_s^2} \end{pmatrix}}_{\sigma(X_s)} dW_s.$$

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with  $\mathcal{D} = \mathcal{D}^1 \times \mathbb{R}_+$ .

Take  $x^2 = 0$ . We want to kill the irregular part :

$$\begin{aligned} 0 &\geq \mathbb{E}_{\mathcal{F}_T^W} \left[ \int_0^t \mathcal{L}\phi(X_s) ds + \int_0^t D\phi(X_s) \sigma(X_s) dW_s \right] \\ &= \int_0^t \mathbb{E}_{\mathcal{F}_T^W} [\mathcal{L}\phi(X_s)] ds + \int_0^t \mathbb{E}_{\mathcal{F}_T^W} [D_1\phi(X_s) \sigma^1(X_s)] dW_s^1 \end{aligned}$$



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and obtain (by the same arguments as before)

$$0 \geq \mathcal{L}\phi(x) - \frac{1}{2} D_1(D_1\phi\sigma^1)(x)\sigma^1(x) = \langle u, b(x) - \frac{1}{2}\sigma^1(x)D_1\sigma^1(x) \rangle.$$

## The case of a regular spectral decomposition

Assume that  $\sigma\sigma^\top$  can be extended into a  $C_{\text{loc}}^{1,1}$  function :

$$C = Q \text{diag} [\lambda_1, \dots, \lambda_r, 0, \dots, 0] Q^\top$$

with  $\lambda_1(x) > \lambda_2(x) > \dots > \lambda_r(x) > 0$  and  $Q(x)Q(x)^\top = I_d$ ,  $r \leq d$ .

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Then,  $\bar{\sigma} : y \mapsto \bar{Q}(y)\bar{\Lambda}(y)^{\frac{1}{2}}$  is  $C^{1,1}(N(x))$ , in which  $\bar{Q} := [q_1 \dots q_r \ 0 \dots 0]$  and  $\bar{\Lambda} = \text{diag}[\lambda_1, \dots, \lambda_r, 0, \dots, 0]$ .

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Moreover,

$$\begin{aligned} 0 &\geq \int_0^t \mathcal{L}\phi(X_s) ds + \int_0^t D\phi(X_s)(Q\Lambda^{\frac{1}{2}}Q^\top)(X_s) dW_s \\ &= \int_0^t \mathcal{L}\phi(X_s) ds + \int_0^t D\phi(X_s)(Q\Lambda^{\frac{1}{2}})(X_s) d\bar{W}_s \end{aligned}$$

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and take expectation given  $\sigma((\bar{W}_s^1, \dots, \bar{W}_s^r), s \leq T)$  to get as above

$$0 \geq \langle u, b(x) - \frac{1}{2} \sum_{j=1}^d DC^j(x)(CC^+)^j(x) \rangle.$$

## The general case

Take  $A_\varepsilon = Q(x)\text{diag}[1 - \varepsilon, (1 - \varepsilon)^2, \dots, (1 - \varepsilon)^d]Q(x)^\top$  so that

$$C_\varepsilon(x) = Q(x)\text{diag}[(1 - \varepsilon)\lambda_1(x), (1 - \varepsilon)^2\lambda_2(x), \dots, (1 - \varepsilon)^d\lambda_d(x)]Q(x)^\top$$

has distinct non-zero eigenvalues and one can apply the above to

$$X_\varepsilon := A_\varepsilon X = A_\varepsilon x + \int_0^\cdot b_\varepsilon(X_s^\varepsilon) ds + \int_0^\cdot C_\varepsilon(X_s^\varepsilon)^{\frac{1}{2}} dW_s$$

with respect to  $\mathcal{D}_\varepsilon := A_\varepsilon \mathcal{D}$ . Then,  $\varepsilon \rightarrow 0$ .

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**Thm :** Assume that  $\sigma\sigma^\top = C$  on  $\mathcal{D}$  for some  $C \in C_{\text{loc}}^{1,1}$ . Then,  $\mathcal{D}$  is stochastically invariant if and only if

$$\begin{cases} C(x)u = 0 \\ \langle u, b(x) - \frac{1}{2} \sum_{j=1}^d DC^j(x)(CC^+)^j(x) \rangle \leq 0 \end{cases}$$

for every  $x \in \mathcal{D}$  and for all  $u \in \mathcal{N}_\mathcal{D}^1(x)$ .

# Extension to jump diffusions by E. Abi Jaber [1]

For the diffusion with jumps

$$X = x + \int_0^\cdot b(X_s) ds + \int_0^\cdot \sigma(X_s) dW_s + \int_0^\cdot \int \rho(X_{s-}, z) (\mu(ds, dz) - F(dz) ds),$$

the conditions become

$$\begin{cases} x + \rho(x, z) \in \mathcal{D}, \text{ for } F\text{-almost all } z, \\ \int |\langle u, \rho(x, z) \rangle| F(dz) < \infty, \\ \sigma(x)^\top u = 0, \\ \langle u, b(x) - \int \rho(x, z) F(dz) - \frac{1}{2} \sum_{j=1}^d DC^j(x) (CC^+)^j(x) \rangle \leq 0, \end{cases}$$

for all  $x \in \mathcal{D}$  and  $u \in \mathcal{N}_{\mathcal{D}}(x)$



An example  
Polynomial diffusions on parabolic concave state space

# Polynomial diffusions

**Definition :**  $X$  is a polynomial diffusion on  $\mathcal{D}$  if :

- (i) There exist  $\bar{b}^i, \tilde{b}^i \in \mathbb{R}$ ,  $0 \leq i \leq 2$ , and  $A^i \in \mathbb{S}^2$ ,  $1 \leq i \leq 5$ , such that  $b : x \mapsto b(x) := (\bar{b}(x), \tilde{b}(x)) \in \mathbb{R}^2$  and  $C : x \mapsto C(x) \in \mathbb{S}^2$  have the following form :

$$\begin{cases} \bar{b}(x) &= \bar{b}^0 + \bar{b}^1 \bar{x} + \bar{b}^2 \tilde{x}, \\ \tilde{b}(x) &= \tilde{b}^0 + \tilde{b}^1 \bar{x} + \tilde{b}^2 \tilde{x}, \\ C(x) &= A^0 + A^1 \bar{x} + A^2 \tilde{x} + A^3 \bar{x}^2 + A^4 \bar{x} \tilde{x} + A^5 \tilde{x}^2, \end{cases}$$

for all  $x = (\bar{x}, \tilde{x}) \in \mathcal{D}$ .

- (ii)  $C(x) \in \mathbb{S}_+^d$ , for all  $x \in \mathcal{D}$ .

When  $A^i = 0$  for all  $3 \leq i \leq 5$ , we say that  $X$  is an affine diffusion.

## Parabolic concave state space

We consider :

$$\mathcal{D} = \{(\bar{x}, \tilde{x}) \in \mathbb{R}^2, \tilde{x} \geq -\bar{x}^2\}.$$

Our conditions are equivalent to

$$\begin{cases} C(x) = C_{11}(x) \begin{pmatrix} 1 & -2\bar{x} \\ -2\bar{x} & 4\bar{x}^2 \end{pmatrix}, \\ \langle u, b(x) \rangle - \frac{\mathbf{1}_{\{C_{11}(x) \neq 0\}}}{2(4\bar{x}^2+1)} [2\bar{x}\partial_u(C_{11} - C_{22})(x) + (1 - 4\bar{x}^2)\partial_u C_{12}(x)] \geq 0, \end{cases}$$

for all  $\bar{x} \in \mathbb{R}$ ,  $x = (\bar{x}, -\bar{x}^2)$  and  $u = (2\bar{x}, 1)^\top \in -\mathcal{N}_{\mathcal{D}}^1(x)$ .

# Necessary and sufficient conditions

- No affine solution unless it has no diffusion part or leaves on the boundary !

## Necessary and sufficient conditions

□ No affine solution unless it has no diffusion part or leaves on the boundary !

□ For polynomial diffusions,  $\mathcal{D}$  is invariant if and only if there exist  $\alpha, \beta \geq 0$  such that either one the following conditions holds :

(a)

$$C(x) = \begin{pmatrix} \alpha & -2\alpha\bar{x} \\ -2\alpha\bar{x} & (4\alpha + \beta)\bar{x}^2 + \beta\tilde{x} \end{pmatrix}, \text{ for all } x = (\bar{x}, \tilde{x}) \in \mathcal{D},$$

(b)  $\bar{b}^2 = 0$  and

$$\begin{cases} \tilde{b}^2 < 2\bar{b}^1 & \text{and} & (\tilde{b}^1 + 2\bar{b}^0)^2 \leq 4(-\tilde{b}^2 + 2\bar{b}^1)(\tilde{b}^0 + \alpha) \\ \text{or} \\ \tilde{b}^2 = 2\bar{b}^1, & \tilde{b}^1 = -2\bar{b}^0 & \text{and} & \tilde{b}^0 \geq -\alpha. \end{cases}$$

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