

# Hamilton-Jacobi approach for some stochastic control problems with state constraints

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# Outline

- 1 Introduction
- 2 Swing options in energy market
- 3 State-constrained stochastic problems: general setting
  - Step 1: reachability problem using unbounded controls
  - Step 2: level set approach
  - Step 3: HJB equation and uniqueness result

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## 1 Introduction

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# Introduction - Motivation

Stochastic differential equations. State and control variables

Let  $(\Omega, \mathbb{F}, \{\mathbb{F}_t\}_{t \geq 0}, \mathbb{P})$  be a filtered probability space. Consider the following controlled SDE in  $\mathbb{R}^d$ :

$$\begin{cases} dX(s) = b(X(s), u(s))ds + \sigma(X(s), u(s))dB_s & s \in [t, T] \\ X(t) = x \end{cases} \quad (1)$$

where  $B$  is a Brownian motion and

$$u \in \mathcal{U} := \{\text{Progr. meas. processes with values in } U\},$$

for a given compact set  $U$ .

$\rightsquigarrow \exists X(\cdot) := X_{t,x}^u(\cdot)$ : unique strong solution of (1) associated to the control  $u$  (under standard regularity assumptions on  $b, \sigma$ ).

# Introduction - Setting of the problem

## State-constrained OCPs

- Let  $\psi : \mathbb{R}^d \rightarrow \mathbb{R}$  (terminal cost) and  $\ell : [0, T] \times \mathbb{R}^d \times U \rightarrow \mathbb{R}$  (running cost).
- Let  $\mathcal{K} \subseteq \mathbb{R}^d$  be a non empty and closed set.

## STATE-CONSTRAINED OCP:

$$\vartheta(t, x) = \inf_{u(\cdot) \in \mathcal{U}} \left\{ \mathbb{E} \left[ \psi(X_{t,x}^u(T)) + \int_t^T \ell(X_{t,x}^u(s), u(s)) ds \right] \text{ s. t. } \right. \\ \left. X_{t,x}^u(s) \in \mathcal{K}, \forall s \in [t, T] \text{ a.s.} \right\}.$$

A particular case: No control,

$$\vartheta(t, x) = \begin{cases} \mathbb{E} \left[ \psi(X_{t,x}(\cdot)) \right] & \text{if } X_{t,x}(\cdot) \subset \mathcal{K} \text{ a.s.,} \\ +\infty & \text{otherwise.} \end{cases}$$

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## Deterministic case: without state constraints

$$\vartheta(t, x) = \inf_{u \in \mathcal{U}} \psi(X_{t,x}^u(T)) + \int_t^T \ell(X_{t,x}^u(s), u(s)) ds$$

- Dynamic Programming Principle (DPP): for  $0 < h \leq T - t$

$$\vartheta(t, x) = \inf_{u \in \mathcal{U}} \left[ \vartheta(t+h, X_{t,x}^u(t+h)) + \int_t^{t+h} \ell(X_{t,x}^u(s), u(s)) ds \right].$$

- The HJB equation associated to  $\vartheta$  is:

$$\begin{aligned} -\partial_t \vartheta(t, x) + H(x, D\vartheta) &= 0 \quad \text{on } (0, T) \times \mathbb{R}^d, \\ \vartheta(T, x) &= \psi(x). \end{aligned}$$

$$\text{with } H(x, p) := \sup_{u \in U} (-b(x, u) \cdot p - \ell(x, u)).$$

- The HJB equation describes a "Front Propagation" phenomena.

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## Stochastic case: without state constraints

$$\vartheta(t, x) = \inf_{u \in \mathcal{U}} \mathbb{E} \left[ \psi(X_{t,x}^u(T)) + \int_t^T \ell(X^u(s), u(s)) ds \right]$$

- The DPP reads as (for  $0 < h \leq T - t$ )

$$\vartheta(t, x) = \inf_{u \in \mathcal{U}} \mathbb{E} \left[ \vartheta(t+h, X_{t,x}^u(t+h)) + \int_t^{t+h} \ell(X_{t,x}^u(s), u(s)) ds \right]$$

- and the HJB equation is

$$-\partial_t \vartheta + H(x, D\vartheta, D^2\vartheta) = 0$$

$$\vartheta(T, x) = \psi(x)$$

$$\text{with } H(x, p, Q) := \sup_{u \in U} \left( -b(x, u) \cdot p - \ell(x, u) - \frac{1}{2} \text{Tr}(\sigma \sigma^T Q) \right).$$

# Viscosity Notion

- The notion of viscosity solution can be used to give a rigorous sense to the solutions of HJB equations
- In many cases the value function is merely continuous or Lipschitz continuous (no more)
- The viscosity notion provides a convenient framework for numerical analysis and error estimates theory. For a consistent, stable and monotone scheme, we get:  $\|\vartheta - \vartheta^h\| = O(h^{1/2})$ .
- Once the value function  $\vartheta$  is known, the optimal control strategy can be computed as:

$$u^*(t, x) \in \operatorname{argmax} H(x, D\vartheta(t, x), D^2\vartheta(t, x)).$$

## Stochastic case: A simple example

- Consider the uncontrolled case ( $\dim=1$ ):

$$\begin{cases} dX(s) = b(X(s))ds + \sigma(X(s))dB_s & s \in [t, T] \\ X(t) = x \end{cases}$$

- Define the value function by:

$$\vartheta_{BS}(t, x) = \mathbb{E}[\psi(X_{t,x}(T))].$$

- The HJB equation associated to  $\vartheta$  (Black & Scholes eq.):

$$\begin{aligned} -\partial_t \vartheta_{BS} - b(x) \cdot D\vartheta_{BS} - \frac{\sigma(x)^2}{2} \Delta \vartheta_{BS} &= 0 \\ \vartheta_{BS}(T, x) &= \psi(x) \end{aligned}$$

In the sequel, we would like to describe the value function  $\vartheta$  in presence of constraints and obstacles.

# Applications

- aircraft landing problem with wind disturbances
- energy consumption/demand problem with limited resource
- super replication problems in finance
- ... any stochastic problem with state constraints

Characterize  $\mathcal{V}$  and design numerical methods for its approximation

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# Swing options in energy market

- Assume  $Y_{t,y}(\cdot)$  is the "price of energy" satisfying:

$$dY_{t,y}(s) = b(Y_{t,y}(s))ds + \sigma(Y_{t,y}(s))dB_s, \quad Y_{t,y}(t) = y$$

- Let  $T > 0$  be the time maturity of the contract, and  $K$  is a pre-specified strike price.
- At every  $s \in [0, T]$ , possibility to buy an amount  $u(s) \in [0, \bar{u}]$  at price  $K$ .
- The cost functional to be maximized:

$$J(Y_{t,y}) = \int_t^T u(s)(Y_{t,x}(s) - K) ds.$$

- Limited resources:

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- **Limited resources:**

$$0 \leq \zeta_{t,z} := z + \int_t^T u(s) ds \leq M.$$

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$$J(Y_{t,y}) = \int_t^T u(s)(Y_{t,x}(s) - K) ds.$$

- **Limited resources:**

$$d\zeta_{t,z}(s) = u(s)ds, \quad \zeta_{t,z}(t) = z,$$

$$0 \leq \zeta_{t,z}(s) \leq M \quad \forall s \in [t, T].$$

# Swing options in energy market

- The state  $(Y_{t,y}(\cdot), \zeta(\cdot))$  satisfies:

$$\begin{aligned}dY_{t,y}(s) &= b(Y_{t,y}(s)) + \sigma(Y_{t,y}(s))dB_s, & Y_{t,y}(t) &= y \\d\zeta_{t,z}(s) &= u(s), & \zeta_{t,z}(t) &= z.\end{aligned}$$

- The control variable:  $u(s) \in [0, \bar{u}]$ .
- State constraints:  $0 \leq \zeta_{t,z}(s) \leq M$  for every  $s \in [t, T]$ .
- Formulation of the control problem (for  $x = (y, z)$ ): .

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- Formulation of the control problem (for  $x = (y, z)$ ):

$$v(t, x) = \min_{u \in \mathcal{U}} \{J(Y_{t,y}(T)), 0 \leq \zeta_{t,z}(T) \leq M\}$$

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- Formulation of the control problem (for  $x = (y, z)$ ):

$$\vartheta(t, x) = \min_{u \in \mathcal{U}} \{J(Y_{t,y}(T)), m(s) \leq \zeta_{t,z}(s) \leq M(s)\}.$$

# Swing options in energy market: some references.

## ■ Discrete time:

- O. Bardou, S. Bouthemy, G. Pages. *Optimal quantization for the pricing of swing options*. Appl. Math. Finance, 2009
- C. Barrera-Esteve, F. Bergeret, C. Dossal, E. Gobet, A. Meziou, R. Munos, D. Reboul-Salze. *Numerical methods for the pricing of swing options: a stochastic control approach*. Methodol. Comput. Appl. Probab, 2006.
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- P. Jaillet, E. I. Ronn, S. Tompaidis. *Valuation of Commodity-Based Swing Options*. Management Science, 2004
- ...

## ■ Continuous time

- R. Carmona, N. Touzi. *Optimal multiple stopping and valuation of swing options*. Math. Finance, 2008.
- F. E. Benth, J. Lempa, T. K. Nilssen. *On the optimal exercise of swing options in electricity markets*. The Journal of Energy Markets, 2012.
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# Introduction

## State-constrained OCPs

$$\vartheta(t, x) = \min_{u \in \mathcal{U}} \mathbb{E} \left[ \psi(X_{t,x}^u(T)) \right] \text{ on processes s.t. } X_{t,x}^u(\cdot) \subset \mathcal{K},$$

with

$$\begin{cases} dX(s) = b(s, X(s), u(s))ds + \sigma(s, X(s), u(s))dB_s & s \in [t, T] \\ X(t) = x \end{cases}$$

- **Unconstrained case** ( $\mathcal{K} = \mathbb{R}^d$ ):  $\vartheta$  is the unique continuous viscosity solution of a second order HJB equation;
- **Constrained case**: Non uniqueness issues due to the loss of regularity of the value function.
  - Deterministic case: Soner ('86), Blanc ('97), Frankowska-Vinter ('00), Hermosilla-HZ ('15), Altarovici-Bokanowski-HZ ('13) ....
  - Stochastic case: Lasry-Lions ('89), Katsoulakis ('94), Bouchard-Nutz ('12) ....

# Stochastic case

## Stochastic OCP with state constraints

- Characterisation of viability kernels:  
Aubin- Da Prato (90,98), ...  
Buckdahn, Peng, Quincampoix, Rainer (98)  
Quincampoix-Rainer (05)
- Stochastic target problems  
Soner-Touzi (00), (02), (09)  
Bouchard-Elie-Imbert (10)

⇒ **A strong compatibility condition is required**

- $\forall x, \forall p \neq 0, \exists a \in U, \langle \sigma(x, a), p \rangle = 0$
- invertibility condition on the matrix  $\sigma$
- or other qualification conditions ...

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# State-constrained stochastic problems: general setting

Consider the following controlled SDE in  $\mathbb{R}^d$ :

$$\begin{cases} dX(s) = b(X(s), u(s))ds + \sigma(X(s), u(s))dB_s & s \in [t, T] \\ X(t) = x \end{cases} \quad (2)$$

where

$$u \in \mathcal{U} := \{\text{Progr. meas. processes with values in } U\}.$$

The **STATE-CONSTRAINED OCP** is:

$$\vartheta(t, x) = \inf_{u \in \mathcal{U}} \left\{ \mathbb{E} \left[ \psi(X_{t,x}^u(T)) \right] \mid X_{t,x}^u(\cdot) \subset \mathcal{K} \text{ a.s.} \right\}$$

# State-constrained stochastic problems: general setting

## Assumptions:

- (S1)  $u(\cdot) \in U$  a.s., with  $U \subseteq \mathbb{R}^m$  compact set;
- (S2)  $b, \sigma : \mathbb{R}^d \times U \rightarrow \mathbb{R}^d, \mathbb{R}^{d \times p}$  are continuous functions, Lipschitz continuous w.r.t.  $x$  (unif. in  $u$ ).  
 $\rightsquigarrow X_{t,x}^u(\cdot)$  unique strong solution of (2).
- (S3)  $\psi : \mathbb{R}^d \rightarrow \mathbb{R}$  Lipschitz continuous.
- (S4)  $\psi \geq 0$

# State-constrained stochastic problems: general setting

Step 1: reachability problem using unbounded controls

$$v(t, x) = \inf_{u \in \mathcal{U}} \left\{ \mathbb{E} \left[ \psi(X_{t,x}^u(T)) \right] \mid X_{t,x}^u(\cdot) \subset \mathcal{K} \text{ a.s.} \right\}$$

It is straightforward to obtain :

$$v(t, x) = \inf \left\{ z \in \mathbb{R} \mid \exists u, \mathbb{E} \left[ \psi(X_{t,x}^u(T)) \right] \leq z \right. \\ \left. \text{and } X_{t,x}^u(s) \in \mathcal{K}, \forall s \in [t, T] \text{ a.s.} \right\}$$

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# State-constrained stochastic problems: general setting

Step 1: reachability problem using unbounded controls

Thanks to Ito's representation theorem this is OK up to a martingale:

$\exists \alpha \in \mathcal{A} := L_{\mathbb{F}}^2$  ( $\mathbb{R}^p$ -valued prog. meas. process):

$$\psi(X_{t,x}^u(T)) = \mathbb{E}[\psi(X_{t,x}^u(T))] + \int_t^T \alpha_s \cdot dB_s, \quad \text{a.s.}$$

$$z \geq \mathbb{E} \left[ \psi(X_{t,x}^u(T)) \right]$$

$$\iff \exists \alpha \in \mathcal{A}, \quad z \geq \psi(X_{t,x}^u(T)) - \int_t^T \alpha_s \cdot dB_s, \quad \text{a.s.}$$



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# State-constrained stochastic problems: general setting

Step 1: reachability problem using unbounded controls

Let

$$Z_{t,x,z}^{u,\alpha}(\cdot) := z + \int_t^{\cdot} \alpha_s \cdot dB_s. \quad (3)$$

In particular,

$$\begin{aligned} z \geq \mathbb{E} \left[ \psi(X_{t,x}^u(T)) \right] &\iff \exists \alpha \in \mathcal{A}, \quad Z_{t,x,z}^{\alpha,u}(T) \geq \psi(X_{t,x}^u(T)) \quad \text{a.s.} \\ &\iff \exists \alpha \in \mathcal{A}, \quad (X_{t,x}^u(T), Z_{t,x,z}^{\alpha,u}(T)) \in \text{epi}(\psi) \quad \text{a.s.} \end{aligned}$$

# State-constrained stochastic problems: general setting

Step 1: reachability problem using unbounded controls

⇒ We consider the following **stochastic state-constrained backward reachable set**

$$\mathcal{R}_t^{\psi, \mathcal{K}} := \left\{ (x, z) \in \mathbb{R}^{d+1} : \exists (u, \alpha) \in \mathcal{U} \times \mathcal{A} \text{ such that } \right. \\ \left. (X_{t,x}^u(T), Z_{t,x,z}^{u,\alpha}(T)) \in \text{epi}(\psi) \text{ and } X_{t,x}^u(s) \in \mathcal{K}, \forall s \in [t, T] \text{ a.s.} \right\}.$$

Then

$$\vartheta(t, x) = \inf \left\{ z \in \mathbb{R} : (x, z) \in \mathcal{R}_t^{\psi, \mathcal{K}} \right\}.$$

# State-constrained stochastic problems: general setting

## Step 2: level set approach

Let

$$g_\psi(x, z) := \max(\psi(x) - z, 0) \quad \text{and} \quad g_{\mathcal{K}}(x) := d(x, \mathcal{K}).$$

In particular:

$$g_\psi, g_{\mathcal{K}} \geq 0 \quad \text{and} \quad g_\psi(x, z) = 0 \Leftrightarrow (x, z) \in \text{epi}(\psi), \quad g_{\mathcal{K}}(x) = 0 \Leftrightarrow x \in \mathcal{K}.$$

## AUXILIARY UNCONSTRAINED OCP:

$$w(t, x, z) = \inf_{(u, \alpha) \in \mathcal{U} \times \mathcal{A}} \mathbb{E} \left[ \underbrace{g_\psi(X_{t,x}^u(T), Z_{t,x,z}^{u,\alpha}(T))}_{\equiv \max(\psi(X_{t,x}^u(T)) - Z_{t,x,z}^{u,\alpha}(T), 0)} + \int_t^T \underbrace{g_{\mathcal{K}}(X_{t,x}^u(s))}_{\equiv d_{\mathcal{K}}(X_{t,x}^u(s))} ds \right]$$

# State-constrained stochastic problems: general setting

## Step 2: level set approach

### Proposition

Then

$$\mathcal{R}_t^{\psi, \mathcal{K}} = \{(x, z) \in \mathbb{R}^{d+1} : w(t, x, z) = 0\}$$

In particular,

$$v(t, x) = \inf \left\{ z \in \mathbb{R} : w(t, x, z) = 0 \right\}.$$

- But  $\alpha_S$  is unbounded ...

## Some References (unbounded controls)

- Lasry - Lions (Cras '00) : "A new class of sing. sto. cont. pbs."  
( $b(x, u) = b_1(x) + b_2(x)u$ ).
- Pham (Prob. Surveys '05) : HJB equation for unbounded controls
- Brüder (Preprint HAL '05) : comparison principle for a super-replication problem in Finance
- Bokanowski- Brüder-Maroso-HZ (SINUM '09) : convergence of SL scheme for a super-replication problem
- Debrabant-Jakobsen (Math. Comp. '13) : high order SL schemes

# State-constrained stochastic problems: general setting

## Step 3: HJB equation and uniqueness result

Recall that

$$\begin{pmatrix} dX_s \\ dZ_s \end{pmatrix} = \begin{pmatrix} b(X_s, u(s)) \\ 0 \end{pmatrix} ds + \begin{pmatrix} \sigma(X_s, u(s)) \\ \alpha \end{pmatrix} dW_s$$

and

$$w(t, x, z) = \inf_{(u, \alpha) \in \mathcal{U} \times \mathcal{A}} \mathbb{E} \left[ \max(\psi(X_{t,x}^u(T)) - Z_{t,x,z}^{u,\alpha}(T), 0) + \int_t^T d_{\mathcal{K}}(X_{t,x}^u(s)) \right]$$

The HJB equation associated to this AUXILIARY OCP would be, in the case  $p = d = 1$ :

$$-\partial_t w + \underbrace{\sup_{\substack{u \in \mathcal{U}, \\ \alpha \in \mathbb{R}}} \left\{ -b \partial_x w - \frac{1}{2} \sigma^2 \partial_{xx} w - \alpha \sigma \partial_{xz} w - \frac{1}{2} \alpha^2 \partial_{zz} w - d_{\mathcal{K}}(x) \right\}}_{=: H(x, z, Dw, D^2w)} = 0$$

⇒ Because of unbounded controls,  $H$  can be unbounded !

# State-constrained stochastic problems: general setting

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⇒ Because of unbounded controls,  $H$  can be unbounded !



# State-constrained stochastic problems: general setting

Interpretation of the eq.  $\tilde{H}(x, z, \partial_t w, Dw, D^2w) = 0$

Notice that (if  $d = p = 1$ )

$$\sup_{\alpha \in \mathbb{R}} \left( A - 2B\alpha + C\alpha^2 \right) = 0 \Leftrightarrow A \leq 0, AC \leq B^2$$

Furthermore,

$$\sup_{\alpha \in \mathbb{R}} A - 2B\alpha + C\alpha^2 = 0$$

$$\Leftrightarrow \sup_{\alpha_1^2 + \alpha_2^2 = 1, \alpha_1 \neq 0} A - 2B \frac{\alpha_2}{\alpha_1} + C \left( \frac{\alpha_2}{\alpha_1} \right)^2 = 0$$

$$\text{"} \Leftrightarrow \text{"} \sup_{\alpha_1^2 + \alpha_2^2 = 1} A\alpha_1^2 - 2B\alpha_1\alpha_2 + C\alpha_2^2 = 0$$

$$\Leftrightarrow \sup_{(\alpha_1, \alpha_2) \in S^1} \begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix}^T \begin{pmatrix} A & -B \\ -B & C \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \end{pmatrix} = 0$$

$$\Leftrightarrow \Lambda_+ \begin{pmatrix} A & -B \\ -B & C \end{pmatrix} = 0$$

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# State-constrained stochastic problems: general setting

Interpretation of the eq.  $\tilde{H}(x, z, \partial_t w, Dw, D^2 w) = 0$

Define the matrix :

$$\mathcal{L}^u(x, z, \partial_t, Dw, D^2 w) := \begin{pmatrix} A & -B \\ -B & C \end{pmatrix}$$

where

$$\frac{A}{2} := -\partial_t w - b \cdot D_x w - \frac{1}{2} \text{Tr}(\sigma \sigma^T D_x^2 w) - d(x, \mathcal{K})$$

$$B := \sigma \partial_{xz} w, \quad \text{and} \quad C := -\partial_{zz} w$$

By elementary calculus

$$\tilde{H}(x, z, w_t, w, D^2 w) = 0$$

$$\Leftrightarrow \sup_{\substack{u \in U \\ \alpha_1^2 + \alpha_2^2 = 1}} \left\{ \alpha_1^2 \left( -\partial_t w - b \partial_x w - \frac{1}{2} \sigma^2 \partial_{xx} w - d(x, \mathcal{K}) \right) - \alpha_1 \alpha_2 \sigma \partial_{xz} w - \frac{1}{2} \alpha_2^2 \partial_{zz} w \right\} = 0$$

$$\Leftrightarrow \sup_{u \in U} \Lambda^+ \left( \mathcal{L}^u(x, z, \partial_t w, Dw, D^2 w) \right) = 0.$$

# State-constrained stochastic problems: general setting

Interpretation of the eq.  $\tilde{H}(x, z, \partial_t w, Dw, D^2 w) = 0$ , for  $d \geq 1$ ,  $p \geq 1$

Define the matrix :

$$\mathcal{L}^u(x, z, \partial_t w, Dw, D^2 w) := \begin{pmatrix} A & -B_1 & \dots & -B_p \\ -B_1 & C & & 0 \\ \vdots & & \ddots & \vdots \\ -B_p & 0 & \dots & C \end{pmatrix}$$

where

$$\begin{aligned} \frac{A}{2} &:= -\partial_t w - b \cdot D_x w - \frac{1}{2} \text{Tr}(\sigma \sigma^T D_x^2 w) - d(x, \mathcal{K}). \\ B &:= \sigma^T D_x \partial_z w = (B_1, \dots, B_p)^T, \quad \text{and} \quad C := -\partial_{zz} w \end{aligned}$$

By elementary calculus:

$$\begin{aligned} \tilde{H}(x, z, w_t, Dw, D^2 w) &= 0 \\ \Leftrightarrow \sup_{u \in \mathcal{U}} \Lambda^+ \left( \mathcal{L}^u(x, z, \partial_t w, Dw, D^2 w) \right) &= 0. \end{aligned}$$

# State-constrained stochastic problems: general setting

## Step 3: HJB equation and uniqueness result

### Theorem

*The auxiliary value function  $w$  is a viscosity solution of the following generalized HJB equation*

$$\begin{cases} \sup_{u \in U} \Lambda^+ \left( \mathcal{L}^u(x, z, Dw, D^2w) \right) = 0, & t < T, (x, z) \in \mathbb{R}^{d+1} \\ w(T, x, z) = g_\psi(x, z) & \mathbb{R}^{d+1} \end{cases}$$

*in the class of continuous function with linear growth at infinity  
( $|w(t, x, z)| \leq C(1 + |x| + |z|)$ )*

**What about uniqueness ??**

## Uniqueness issue: the uncontrolled case.

Consider the solution to the HJB equation:

$$\Lambda^+ \left( \begin{array}{ccc} -\frac{\partial \vartheta}{\partial t} - b(t, x) \frac{\partial \vartheta}{\partial y} - \frac{1}{2} \sigma^2(t, x) \frac{\partial^2 \vartheta}{\partial x^2} & -\frac{1}{2} \sigma(t, x) \frac{\partial^2 \vartheta}{\partial x \partial z} \\ -\frac{1}{2} \sigma(t, x) \frac{\partial^2 \vartheta}{\partial x \partial z} & -\frac{1}{2} \frac{\partial^2 \vartheta}{\partial z^2} \end{array} \right) = 0$$

In particular,  $\vartheta$  is a viscosity super-solution of:

$$-\frac{\partial^2 \vartheta}{\partial z^2} \geq 0.$$

$\Lambda^+(J)$  denotes the biggest eigenvalue of  $J$ .

$$\Lambda^+(M) = 0 \Leftrightarrow \max_{\|\alpha\|=1} \alpha^T M \alpha = 0.$$

The HJB equation can be rewritten:

$$\max_{\alpha_1^2 + \alpha_2^2 = 1} \left\{ -\alpha_1^2 \frac{\partial v}{\partial t} - \alpha_1^2 b(t, x) \frac{\partial v}{\partial x} - \frac{1}{2} \text{tr}[a(t, x, \alpha) D^2 v] \right\} = 0.$$

$$a(t, x, \alpha) := \begin{pmatrix} \sigma^2(t, x) \alpha_1^2 & \alpha_1 \alpha_2 \sigma(t, x) \\ \alpha_1 \alpha_2 \sigma(t, x) & \alpha_2^2 \end{pmatrix}.$$

**Remark:**  $\alpha_1$  could vanish.

- If  $\alpha_1 = 0 \Rightarrow$  Elliptic equation.
- If  $\alpha_1 \neq 0 \Rightarrow$  Parabolic equation.

$$\Lambda^+(M) = 0 \Leftrightarrow \max_{\|\alpha\|=1} \alpha^T M \alpha = 0.$$

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$$a(t, x, \alpha) := \begin{pmatrix} \sigma(t, x) \alpha_1 \\ \alpha_2 \end{pmatrix} \begin{pmatrix} \sigma(t, x) \alpha_1 \\ \alpha_2 \end{pmatrix}^T$$

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# State-constrained stochastic problems: general setting

Come back to the general case: uniqueness result

**Boundary conditions for  $z \leq 0$ :**

Let  $w_0$  be the value of the following unconstrained problem:

$$w_0(t, x) := \inf_{u \in \mathcal{U}} \mathbb{E} \left[ \psi(X_{t,x}^u(T)) + \int_t^T d(X_{t,x}^u(s), \mathcal{K}) ds \right].$$

Proposition (A Dirichlet boundary condition)

For any  $z \leq 0$ ,  $w(t, x, z) \equiv w_0(t, x) - z$ .

*Proof:* For any  $z \in \mathbb{R}$ ,

$$\begin{aligned} w(t, x, z) &\geq \inf_{(u, \alpha) \in \mathcal{U} \times \mathcal{A}} \mathbb{E} \left[ \psi(X_{t,x}^u(T)) - z - \underbrace{\int_t^T \alpha_s dB_s}_{\mathbb{E}(\cdot) \equiv 0} + \int_t^T g_{\mathcal{K}}(X_{t,x}^u(s)) ds \right] \\ &\geq w_0(t, x) - z \end{aligned}$$

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# State-constrained stochastic problems: general setting

## Step 3: HJB equation and uniqueness result

*Proof (continued):* On the other hand, for  $z \leq 0$ , choosing the particular control  $\alpha_s := 0$ :

$$\begin{aligned} w(t, x, z) &\leq \inf_{u \in \mathcal{U}} \mathbb{E} \left[ \underbrace{\max(\psi(X_{t,x}^u(T)) - z, 0)}_{\geq 0} + \int_t^T g_{\mathcal{K}}(\cdot) ds \right] \\ &\leq w_0(t, x) - z \end{aligned}$$

# State-constrained stochastic problems: general setting

## Step 3: HJB equation and uniqueness result

### Theorem (Comparison principle, Dirichlet case (Bokanowski-Picarelli-HZ'15))

*The value function is the unique viscosity solution to the following HJB equation*

$$\sup_{u \in U} \Lambda^+ \left( \mathcal{L}^u(x, z, \partial_t w, Dw, D^2 w) \right) = 0, \quad t < T, x \in \mathbb{R}^d, z \geq 0,$$

$$w(T, x, z) = g_\psi(x, z) \equiv \max(\psi(x) - z, 0), \quad x \in \mathbb{R}^d, z \geq 0,$$

$$w(t, x, 0) = w_0(t, x), \quad t < T, x \in \mathbb{R}^d$$

*w has linear growth in x ( $|w| \leq C(1 + |x|)$ )*

# Conclusion

- state-constrained OCP can be recasted into a state-constrained reachability problem, adding
  - a state variable
  - an  $\mathbb{R}^p$ -valued unbounded control
- the state-constrained reachability problem can be modelled by a level set approach and an auxiliary unconstrained OCP
- The value of this OCP is characterized as the unique solution of a "special" HJB equation. It is continuous and can be approximated by a large panel of numerical methods (finite differences, Semi-Lagrangian methods, Markov Chain approximations, ...).

Thanks for your attention!!