

# Lower semicontinuous value functions for target control problems with state constraints

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# Outline

- 1 Introduction
  - Known results on control problem with state constraints
  - Minimum time problem with state constraints
- 2 The HJB equation associated to our problem
  - Geometric approach. Contingent derivatives
  - A second approach. Viscosity notion
- 3 NUmerical test
- 4 Numerical example

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# Control problem with state constraints

$$v(T, \mathbf{x}) := \text{Min } \varphi(\mathbf{y}_x(T));$$

$$\begin{cases} \dot{\mathbf{y}}_x(t) = f(\mathbf{y}_x(t), u(t)), & t > 0, \\ \mathbf{y}_x(0) = \mathbf{x}, \end{cases} \quad (1a)$$

$$u(t) \in \mathcal{U} := L^\infty(0, +\infty; U) \text{ a.e. } t > 0 \quad (1b)$$

$$\mathbf{y}_x(T) \in \mathcal{C}, \quad \mathbf{y}_x(t) \in \mathcal{K} \text{ for } 0 \leq t \leq T. \quad (1c)$$

$\mathcal{C} \subset \mathcal{K}$  closed sets of  $\mathbb{R}^d$  ( $\mathcal{C} \neq \emptyset$ ),  $U \subset \mathbb{R}^m$  a compact set, the functions  $f : \mathbb{R}^d \times U \rightarrow \mathbb{R}^d$  and  $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$  are Lipschitz, bounded, and  $f(x, U)$  is convex for any  $x \in \mathbb{R}^d$ .

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# Free state constraints ( $\mathcal{K} = \mathcal{C} = \mathbb{R}^d$ )

$v$  is the unique bounded continuous viscosity solution of the HJB equation:

$$\begin{aligned} \partial_t V - H(x, D_x V(x)) &= 0 && \text{in } x \in \mathbb{R}^d, t > 0 \\ V(x, 0) &= \varphi(x) && x \in \mathbb{R}^d, \end{aligned}$$

where  $H(x, D_x V(x)) = \min_{a \in U} \left( f(x, a) \cdot D_x V(x) \right)$ .

(Lions, Barles, Cappuzzo-Dolcetta, ... etc)

# RDV problem ( $\mathcal{K} = \mathbb{R}^d$ , $\mathcal{C} \subset \mathbb{R}^d$ )

In this case, the value function  $v$  is the unique l.s.c. 'Bilateral' solution

$$\begin{aligned} \partial_t V - H(x, D_x V(x)) &= 0 && \text{in } ]0, T[ \times \mathbb{R}^d, \\ V(x, 0) &= \varphi(x) \chi_{\mathcal{C}}(x) && x \in \mathbb{R}^d, \end{aligned}$$

(Barron-Jensen, Frankowska, ...)

# RDV problem ( $\mathcal{K} = \mathbb{R}^d$ , $\mathcal{C} \subset \mathbb{R}^d$ )

Bilateral solution: **touching one side test function**

## Definition (Barron-Jensen, Frankowska)

A l.s.c. function  $v$  is a bilateral solution of the HJB equation, if

- (i) for every  $\phi \in C^1$  s. t.  $u - \phi$  has a local minimum at  $(x, t) \in \mathbb{R}^d \times \mathbb{R}_*^+$ ,

$$\partial_t \phi(x, t) - H(x, D_x \phi(x, t)) = 0,$$

- (ii)  $v(x, 0) = \varphi(x) \chi_{\mathcal{C}}(x) = \liminf_{\substack{y \rightarrow x \\ t \searrow 0^+}} v(y, t),$

$$\forall x \in \mathbb{R}^d$$



# Known results

State constraints ( $\mathcal{K} = \mathcal{C}$ )

- **Inward qualification constraint:**

$$\min_{a \in U} f(x, a) \cdot \eta_x < 0, \quad \forall x \in \partial \mathcal{K}$$

$v$  is continuous on  $\mathcal{K}$  (Ishii-Koike, Soner, ...)

- **Outward qualification constraint:**

$$\sup_{a \in U} f(x, a) \cdot \eta_x > 0, \quad \forall x \in \partial \mathcal{K}$$

The value function is the unique l.s.c. viscosity solution of the HJB equation on  $\text{int}(\mathcal{K})$  and supersolution on  $\partial \mathcal{K}$ .

(Frankowska, Vinter, Plaskasz)

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# Target problem with state constraints ( $\mathcal{P}$ )

$$\begin{aligned} \vartheta(T, \mathbf{x}) &:= \text{Min } \Psi(y_{\mathbf{x}}(\tau)); \\ &\begin{cases} \dot{y}_{\mathbf{x}}(t) = f(y_{\mathbf{x}}(t), u(t)), & t > 0, \\ y_{\mathbf{x}}(0) = \mathbf{x}, \end{cases} \\ &u(t) \in \mathcal{U} := L^\infty(0, +\infty; U) \text{ a.e. } t > 0, \\ &\tau \in [0, T], \quad y_{\mathbf{x}}(\tau) \in \mathcal{C}, \\ &y_{\mathbf{x}}(t) \in \mathcal{K} \text{ for } 0 \leq t \leq \tau. \end{aligned}$$

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$$y_{\mathbf{x}}(t) \in \mathcal{K} \text{ for } 0 \leq t \leq \tau.$$

$\mathcal{C} \subset \mathcal{K}$  closed sets of  $\mathbb{R}^d$ ,  $U \subset \mathbb{R}^m$  compact,  
 $f : \mathbb{R}^d \times U \rightarrow \mathbb{R}^d$  is Lipschitz, bounded, and  $f(x, U)$  is  
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$$\Psi(\mathbf{x}) = \begin{cases} 0 & \mathbf{x} \in \mathcal{C} \\ 1 & \text{otherwise} \end{cases} = \chi_{\mathcal{C}}(\mathbf{x}).$$

- **Link with the optimal time problem**

$$\mathcal{T}(\mathbf{x}) := \inf\{\tau \geq 0 \mid y_{\mathbf{x}}(\tau) \in \mathcal{C}, y_{\mathbf{x}} \subset \mathcal{K}\}.$$

### Theorem

➤  $\vartheta(T, \mathbf{x}) = 0 \iff \mathcal{T}(\mathbf{x}) \leq T$

➤ For every  $\mathbf{x} \in \mathbb{R}^d$ ,

$$\left[ \vartheta(T, \mathbf{x}) = 1, \forall T \geq 0 \right] \iff \mathcal{T}(\mathbf{x}) = +\infty.$$

➤  $\mathcal{T}(\mathbf{x}) = \inf\{T \geq 0; \vartheta(T, \mathbf{x}) = 0\}.$

- **Link with front propagation:**

The target set  $\mathcal{C}$ : **burned region at  $t = 0$**



- **Link with the optimal time problem**

$$\mathcal{T}(x) := \inf\{\tau \geq 0 \mid y_x(\tau) \in \mathcal{C}, y_x \subset \mathcal{K}\}.$$

- **Link with front propagation:**

The target set  $\mathcal{C}$ : **burned region at  $t = 0$ ,**

$$\vartheta(0, \mathbf{x}) = \chi_{\mathcal{C}}(\mathbf{x}).$$

$\mathcal{C}_t := \{\mathbf{x} \in \mathbb{R}^n, \vartheta(t, \mathbf{x}) = 0\}$ : **burned region at time  $t$ .**

"Evolution of regions" instead of the "level set approach".

$$\begin{cases} \dot{y}_x(t) = f(y_x(t), u(t)) \\ y_x(0) = x \end{cases} \rightsquigarrow \begin{cases} \dot{\tilde{y}}_x(t) = \lambda(t) f(\tilde{y}_x(t), \tilde{u}(t)) \\ \tilde{y}_x(0) = x \\ \lambda(t) \in \Lambda(\tilde{y}_x(t)) \end{cases}$$

$$\Lambda(z) = \begin{cases} \{0\} & \text{if } z \in \mathcal{K}^c \\ [0, 1] & \text{if } z \in \mathcal{C} \cup \partial\mathcal{K} \\ \{1\} & \text{otherwise.} \end{cases}$$

## Lemma 2.

Let  $T \geq 0$ ,  $x \in \mathcal{K}$ . For every  $\tilde{y}_x : [0, T] \rightarrow \mathcal{K}$ , there exist  $0 \leq S \leq T$ ,  $u \in \mathcal{U}$  such that:

$$\begin{aligned} \dot{y}_x(t) &= f(y_x(t), u(t)) \quad t \in (0, S), & y_x(0) &= x, \\ y_x(S) &= \tilde{y}_x(T), \end{aligned}$$

$$\{y_x(t); t \in (0, S)\} \equiv \{\tilde{y}_x(t); t \in [0, T]\} \subset \mathcal{K}.$$

## Idea of the proof of Lemma 2:

$\tilde{y}_x$  satisfies:  $\exists \lambda : (0, T) \longrightarrow [0, 1]$  such that:

$$\begin{cases} \dot{\tilde{y}}_x(t) = \lambda(t)f(\tilde{y}_x(t), \tilde{u}(t)), & \lambda(t) \in \Lambda(\tilde{y}_x(t)). \\ \tilde{y}_x(0) = x \end{cases}$$

Let  $\gamma(t) := \int_0^t \lambda(s)ds$ . We construct  $y$  such that

$$y_x(\gamma(t)) = \tilde{y}_x(\beta(t)); \quad S = \gamma(T);$$

where  $\beta(t) = \inf\{0 \leq \tau \leq t, \gamma(\tau) = \gamma(t)\}$ .

## An other formulation of the target problem

$$v(T, \mathbf{x}) = \inf \{ \chi_{\mathcal{C}}(\tilde{\mathbf{y}}_{\mathbf{x}}(t)), \dot{\tilde{\mathbf{y}}}_{\mathbf{x}}(t) \in F(\tilde{\mathbf{y}}_{\mathbf{x}}(t)), \tilde{\mathbf{y}}_{\mathbf{x}}(0) = \mathbf{x}, \}$$

where  $F(\mathbf{x}) = \{ \lambda f(\mathbf{x}, \mathbf{a}), \mathbf{a} \in \mathbf{U}, \lambda \in \Lambda(\mathbf{x}) \}$  et

$$\Lambda(\mathbf{x}) := \begin{cases} \{0\} & \text{if } \mathbf{x} \in \mathcal{K}^c \\ [0, 1] & \text{if } \mathbf{x} \in \mathcal{C} \cup \partial\mathcal{K} \\ \{1\} & \text{otherwise.} \end{cases}$$

## Question

What is the HJB equation satisfied by  $v$ ?

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## Contingent epiderivative of a l.s.c function $u$ :

$$D_{\uparrow} u(t, x)(s, p) := \liminf_{h \searrow 0, q \rightarrow p} \frac{1}{h} (u(t+hs, x+hq) - u(t, x))$$

When  $F(x) := f(x, U)$  ( $\mathcal{K} = \mathbb{R}^d$ ),  $F$  is cont.,

$$u = v \Leftrightarrow \begin{cases} u \text{ lsc, verifies (IC): } \liminf_{t \searrow 0, y \rightarrow x} u(y, t) = \chi_c(x), \\ \forall t > 0, x \in \mathbb{R}^n, \sup_{p \in F(x)} -D_{\uparrow} u(t, x)(-1, p) \geq 0 \\ \forall t \geq 0, x \in \mathbb{R}^n, \sup_{p \in F(x)} D_{\uparrow} u(t, x)(1, -p) \leq 0. \end{cases}$$



## Contingent epiderivative of a l.s.c function $u$ :

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$$Tr^-(x) := \left\{ \tilde{y}_x \mid \exists \nu > 0 \tilde{y}_x(\cdot) \subset F(\tilde{y}_x(\cdot)), \text{ on } [-\nu, 0], \right. \\ \left. \text{et } \tilde{y}_x(0) = x \right\},$$

with  $F(x) = \{\lambda f(x, a), a \in U, \lambda \in \Lambda(x)\}$

$$D_{\uparrow}^F u(t, x)(\tilde{y}_x) := \liminf_{h \searrow 0} \frac{1}{h} (u(t+h, \tilde{y}_x(-h)) - u(t, x))$$

Derivative with respect to trajectories arriving in  $x$ .

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Derivative with respect to trajectories arriving in  $x$ .

## Theorem (O. Bokanowski, HZ, N. Megdich)

$$u = v \Leftrightarrow \begin{cases} u \text{ is l.s.c, and satisfies the IC,} \\ \forall t > 0, x \in K, \sup_{p \in F(x)} -D_{\uparrow} u(t, x)(-1, p) \geq 0, \\ \forall t \geq 0, x \in K, \sup_{y \in Tr^{-}(x)} D_{\uparrow}^F u(t, x)(y) \leq 0, \end{cases}$$

$$u = \vartheta \Leftrightarrow \left\{ \begin{array}{l} u \text{ is l.s.c., and satisfies IC,} \\ \forall t > 0, x \in K, \sup_{p \in F(x)} -D_{\uparrow} u(t, x)(-1, p) \geq 0 \\ \forall t \geq 0, \forall x \in K, \sup_{y \in Tr^{-}(x)} D_{\uparrow}^F u(t, x)(y) \leq 0, \end{array} \right.$$

$$\Leftrightarrow \left\{ \begin{array}{l} u \text{ is l.s.c., and satisfies IC,} \\ \forall t > 0, x \in K, \sup_{p \in F(x)} -D_{\uparrow} u(t, x)(-1, p) \geq 0, \\ \forall t \geq 0, x \in \text{int}(K), \sup_{p \in F(x)} D_{\uparrow} u(t, x)(1, -p) \leq 0 \\ \forall t \geq 0, \forall x \in \partial K, \sup_{y \in Tr^{-}(x)} D_{\uparrow}^F u(t, x)(y) \leq 0, \end{array} \right.$$

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## Theorem (OB, HZ, NM)

*The value function  $V$  is l.s.c on  $\mathbb{R}^d \times \mathbb{R}_*^+$ , and is a bilateral solution of the HJB equation*

$$\min(\partial_t \vartheta(\mathbf{x}, t) - \mathcal{H}(\mathbf{x}, D_x \vartheta(\mathbf{x}, t)), \vartheta - \chi_{\mathcal{K}}(\mathbf{x})) = 0,$$

$$V(\mathbf{x}, 0) = \Phi(\mathbf{x}) \quad \mathbf{x} \in \mathbb{R}^d,$$

*where  $\mathcal{H}(\mathbf{x}, p) = \inf\{\lambda f(\mathbf{x}, \mathbf{a}); \lambda \in \Lambda(\mathbf{x}), \mathbf{a} \in U\}$ ,*

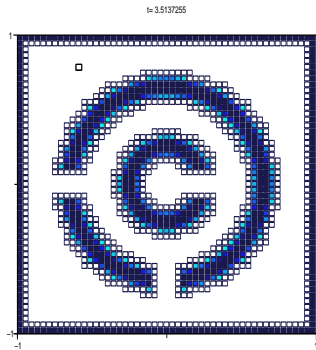
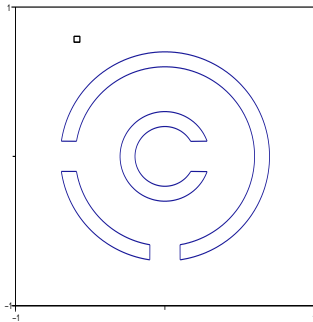


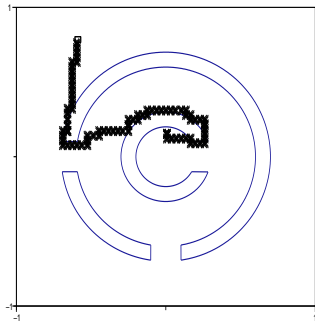
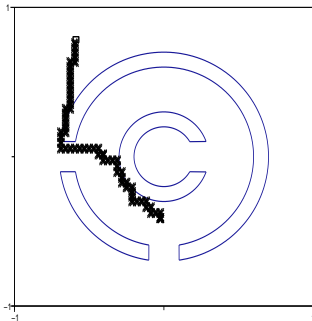
# Algorithme: (without proof)

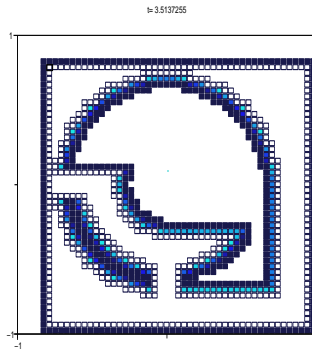
$$\min\left(\frac{V^{n+1} - V^n}{\Delta t} + [H(V^n)], V^{n+1} - \chi_K\right) = 0,$$

$$\Rightarrow V^{n+1} = \max(V^n - \Delta t[H(V^n)], \chi_K)$$

- 2d sparse method - The HJB-UltraBee (Bokanowski, Megdich, Zidani'05)







Introduction

The HJB equation associated to our problem

NUmerical test

**Numerical example**

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