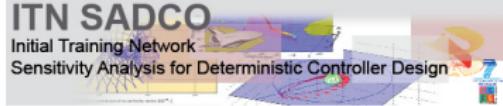


# HJ equations. Reachability analysis. Optimal control problems

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# Outline

- 1 Introduction
- 2 The viscosity Notion
- 3 Controlled systems. Optimal control problems
- 4 Value functions. HJB equations
- 5 Comparison principle
- 6 Parabolic HJB equations. Stochastic control problems
- 7 HJ equation on multidomains. Transmission conditions
- 8 Link between HJB equations and optimal control problems

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## Some steady equations

- $\lambda v(x) + H(x, Dv(x)) = 0 \quad x \in \mathbb{R}^d$
- $$\begin{cases} H(x, Dv(x)) = 0 & x \in \Omega \\ v(x) = 0 & x \in \partial\Omega \end{cases}$$
 where  $\Omega \subset \mathbb{R}^d$

## Time-dependant equations

- $$\begin{cases} \partial_t v(t, x) + H(t, x, Dv(t, x)) = 0 & x \in \mathbb{R}^d, t > 0 \\ v(0, x) = v_0(x) & x \in \mathbb{R}^d \end{cases}$$
- $$\begin{cases} \min(\partial_t v(t, x) + H(t, x, Dv(t, x)), v(t, x) - g(t, x)) = 0 \\ v(0, x) = v_0(x) \end{cases}$$

# Hamiltonian $H$

- $H : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$  is a Lipschitz continuous function
- In the sequel, we will consider the particular case where the Hamiltonian  $H$  is defined by

$$H(x, p) := \sup_{q \in F(x)} (-q \cdot p - \ell(x, q))$$

where  $F : \mathbb{R}^d \rightrightarrows \mathbb{R}^d$  is a multi-valued function, and  
 $\ell : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$ .

## Some examples

- Eikonal equation:

$$\begin{cases} \partial_t v(t, x) + c(x) \|Dv(t, x)\| = 0 & x \in \mathbb{R}^d, t > 0 \\ v(0, x) = v_o(x) & x \in \mathbb{R}^d \end{cases}$$

Here the velocity  $c(\cdot) \geq 0$ , and the Hamiltonian is defined by

$$H(x, p) := c(x) \|p\| = \max_{q \in \mathbb{B}(0, 1)} (-q \cdot p).$$

- Advection equation with an obstacle:

$$\begin{cases} \min (\partial_t v(t, x) - \vec{a}(x) \cdot Dv(t, x), v(t, x) - g(t, x)) = 0 \\ v(0, x) = v_o(x) \quad x \in \mathbb{R}^d \end{cases}$$

the Hamiltonian is defined by  $H(x, p) := \vec{a}(x) \cdot p$ .

# Hamilton-Jacobi-Bellman (HJB for short) equations

- Notion of solutions for HJB equations
- Existence of solutions: Value function of optimal control problems
- Uniqueness results: Comparison principle

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## Definition.

Let  $\Omega \subset \mathbb{R}^M$ , and  $\mathcal{F} : \Omega \times \mathbb{R} \times \mathbb{R}^M \rightarrow \mathbb{R}$  be continuous. Consider the HJB equation:

$$\mathcal{F}(x, u(x), Du(x)) = 0 \quad \text{in } \Omega. \quad (1)$$

(i) A function  $u \in USC(\Omega)$  is a viscosity **sub-solution** of (7) if for any  $\varphi \in C^1(\Omega)$  and any local maximum point  $x_0 \in \Omega$  of  $u - \varphi$ ,

$$\mathcal{F}(x_0, u(x_0), D\varphi(x_0)) \leq 0.$$

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(ii) A function  $u \in LSC(\Omega)$  is a viscosity **super-solution** of (7) if for any  $\varphi \in C^1(\Omega)$  and any local minimum point  $x_0 \in \Omega$  of  $u - \varphi$ ,

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(iii) A function  $u \in C(\Omega)$  is a viscosity **solution** of (7) if it is a sub- and super-solution.

# Vanishing Viscosity Limits

## Lemma

Let  $u_\varepsilon$  be a sequence of smooth solutions to the **viscous** equations:

$$\mathcal{F}(x, u_\varepsilon(x), Du_\varepsilon(x)) - \varepsilon \Delta u_\varepsilon(x) = 0 \quad x \in \Omega.$$

If  $u_\varepsilon \rightarrow u$  uniformly on  $\Omega$ , when  $\varepsilon \rightarrow 0+$ , then  $u$  is a viscosity solution of

$$\mathcal{F}(x, u(x), Du(x)) = 0.$$

## Idea of the proof - 1:

- $u_\varepsilon$  is a viscosity sub-solution to the viscous equation.  
Indeed, let  $\psi \in C^1(\Omega)$ , and let  $x \in \Omega$  a local maximum of  $u_\varepsilon - \psi$ .  
Then

$$Du_\varepsilon(x) = D\psi(x), \quad \Delta u_\varepsilon(x) \leq \Delta\psi(x).$$

Therefore,

$$\mathcal{F}(x, u_\varepsilon(x), Du_\varepsilon(x)) - \varepsilon \Delta u_\varepsilon(x) \leq \mathcal{F}(x, u_\varepsilon(x), D\psi(x)) - \varepsilon \Delta\psi(x) = 0.$$

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Therefore,

$$\mathcal{F}(x, u_\varepsilon(x), D\psi(x)) - \varepsilon\Delta\psi(x) \leq 0.$$

- Let  $\varphi \in C^1(\Omega)$ ,  $x \in \Omega$  is a local maximum of  $u - \varphi$ .  
 $\forall \rho > 0$ ,  $\exists \delta \leq \rho$  and there exists  $\psi \in C^1(\Omega)$  s.t.

$$\|\psi - \varphi\|_{C^1(\Omega)} \leq \delta$$

$u_\varepsilon - \psi$  has a local maximum  $x_\varepsilon$  inside the ball  $B(x, \rho)$ .

## Idea of the proof - 2:

- Extract a convergent subsequence  $x_\varepsilon \rightarrow \bar{x} \in B(x, \rho)$ . By passing to the limit in the inequality

$$\mathcal{F}(x_\varepsilon, u_\varepsilon(x_\varepsilon), D\psi(x_\varepsilon)) - \varepsilon \Delta\psi(x_\varepsilon) \leq 0,$$

we get:

$$\mathcal{F}(\bar{x}, u(\bar{x}), D\psi(\bar{x})) \leq 0.$$

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- Choosing  $\rho$  small enough, we have  $|\bar{x} - x|$  and  $|D\psi(\bar{x}) - D\varphi(x)|$  as small as we want. So by continuity of  $\mathcal{F}$ , we finally obtain:

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- The supersolution can be proved in the same way.

# A more general stability result

- A general stability theorem holds for viscosity solutions, without additional requirements on the derivatives.

## Theorem

Consider a sequence of continuous functions  $(u_n)_{n \geq 1}$ , where  $u_n$  is a viscosity sub-solutions (resp. super-solutions) to

$$\mathcal{F}_n(x, u_n(x), Du_n(x)) = 0, \quad x \in \Omega.$$

Assume  $\mathcal{F}_n \rightarrow \mathcal{F}$  uniformly on compact subsets of  $\Omega \times \mathbb{R}^d \times \mathbb{R}^d$ , when  $n \rightarrow +\infty$  and  $u_n \rightarrow u$  uniformly on  $\Omega$ .

Then  $u$  is a viscosity sub-solution (a super-solution) of

$$\mathcal{F}(x, u(x), Du(x)) = 0, \quad x \in \Omega.$$

- The stability results are based on the following "key ingredient":

### Lemma

Let  $v : \Omega \rightarrow \mathbb{R}$  be a upper semi-continuous function that achieves a *strict* local maximum at  $\bar{x} \in \Omega$ . Let  $(v_n)_n$  be a sequence of upper semi-continuous function on  $\Omega$ .

Assume that  $\limsup_{\substack{z \rightarrow \bar{x} \\ n \rightarrow +\infty}} v_n(z) = v(\bar{x})$ . Then there exists a sequence  $(x_n)_n$

in  $\Omega$  such that for every  $n \geq 1$ ,  $x_n$  is a local maximum of  $v_n$  and

$$\lim_{n \rightarrow +\infty} x_n = \bar{x}, \quad \lim_{n \rightarrow +\infty} v(x_n) = v(\bar{x}).$$

# Semi-relaxed limits

Let  $(u_n)_n$  be a sequence of functions on  $\Omega$ .

- Assume that  $(u_n)_n$  are locally uniformly bounded from above  
Define the *upper relaxed limit* by:

$$u^*(x) := \limsup_{x_n \rightarrow x} u_n(x_n).$$

- Assume that  $(u_n)_n$  are locally uniformly bounded from below.  
Define the *lower relaxed limit* by:

$$u_*(x) := \liminf_{x_n \rightarrow x} u_n(x_n).$$

## Theorem (Stability by semi-relaxed limits)

Consider a sequence of usc (resp. lsc) functions  $(u_n)_{n \geq 1}$ , where  $u_n$  is a viscosity sub-solutions (resp. super-solutions) to

$$\mathcal{F}_n(x, u_n(x), Du_n(x)) = 0, \quad x \in \Omega.$$

Assume  $\mathcal{F}_n \rightarrow \mathcal{F}$  uniformly on compact subsets of  $\Omega \times \mathbb{R}^d \times \mathbb{R}^d$ , when  $n \rightarrow +\infty$  and  $(u_n)_n$  is locally uniformly bounded from above (resp. from below) on  $\Omega$ .

Then  $u^*$  is a viscosity sub-solution (resp  $u_*$  is a super-solution) of

$$\mathcal{F}(x, u(x), Du(x)) = 0, \quad x \in \Omega.$$

## Viscosity notion: An equivalent definition.

Let  $\Omega \subset \mathbb{R}^M$ , and  $\mathcal{F} : \Omega \times \mathbb{R} \times \mathbb{R}^M \rightarrow \mathbb{R}$  be continuous. Consider the HJB equation:

$$\mathcal{F}(x, u(x), Du(x)) = 0 \quad \text{in } \Omega, \tag{2}$$

(i) A function  $u \in USC(\Omega)$  is a viscosity sub-solution of (7) if

$$\mathcal{F}(x, u(x), q) \leq 0 \quad \forall q \in D^+ u(x).$$

$$q \in D^+ u(x) \iff u(y) \leq u(x) + q \cdot (y - x) + o(|y - x|),$$

$$q \in D^- u(x) \iff u(y) \geq u(x) + q \cdot (y - x) + o(|y - x|).$$

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(iii) A function  $u \in C(\Omega)$  is a viscosity solution of (7) if it is a sub- and super-solution.

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$$q \in D^- u(x) \iff u(y) \geq u(x) + q \cdot (y - x) + o(|y - x|).$$

## Lemma

Suppose  $u \in C(\Omega)$ .

- (i)  $q \in D^+ u(x)$  if and only if there exists  $\varphi \in C^1(\Omega)$  such that  $D\varphi(x) = q$  and  $x$  is a local minimum of  $u - \varphi$ .
- (ii)  $q \in D^- u(x)$  if and only if there exists  $\varphi \in C^1(\Omega)$  such that  $D\varphi(x) = q$  and  $x$  is a local maximum of  $u - \varphi$ .

## Some remarks

- ☞ If  $u$  is differentiable at  $x \in \Omega$ , then  $D^- u(x) = D^+ u(x) = \{\nabla u(x)\}$ .
- ☞ A viscosity solution  $u$  of the HJ equation satisfies the equality

$$\mathcal{F}(x, u(x), \nabla u(x)) = 0$$

in the classical sense whenever  $u$  is differentiable at  $x$ .

- ☞ A Lipschitz continuous viscosity solution  $u$  is also a solution in the "*almost everywhere*" sense. (Indeed, by Rademacher's theorem, a Lipschitz continuous function is differentiable a.e.)

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# Controlled Systems

$$\begin{cases} \dot{y}_x(s) \in F(y_x(s)), & s \in (0, 1), \\ y_x(0) = x, \end{cases}$$

► **(H1)**  $F : \mathbb{R}^d \rightsquigarrow \mathbb{R}^d$  is a **usc** set-valued function

$$\forall x_0 \in \mathbb{R}^d, \forall \varepsilon > 0, \exists \eta > 0,$$

$$\|x - x_0\| \leq \eta \implies F(x) \subset F(x_0) + \varepsilon B.$$

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► **(H1)**  $F : \mathbb{R}^d \rightsquigarrow \mathbb{R}^d$  is a **usc** set-valued function, with convex compact nonempty images, and  $F$  has a **linear growth**

$$\exists k > 0, \forall x \in \mathbb{R}^d, \sup_{v \in F(x)} \|v\| \leq k(1 + \|x\|).$$

# Controlled Systems

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- **(H1)**  $F : \mathbb{R}^d \rightsquigarrow \mathbb{R}^d$  is a **usc** set-valued function, with convex compact nonempty images, and  $F$  has a **linear growth**
- **(H2)**  $F : \mathbb{R}^d \rightsquigarrow \mathbb{R}^d$  is Lipschitz continuous  
 $\exists C > 0, \text{s.t. } F(y_2) \subset F(y_1) + C\|y_2 - y_1\|B, \quad \forall y_1, y_2 \in \mathbb{R}^d$

# Controlled Systems

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- **(H2)**  $F : \mathbb{R}^d \rightsquigarrow \mathbb{R}^d$  is Lipschitz continuous
- Example:  $F(x) := f(x, U)$ ; where  $U \subset \mathbb{R}^m$  compact,  
 $f : \mathbb{R}^d \times U \rightarrow \mathbb{R}^d$  is Lipschitz, bounded, and  $f(x, U)$  is convex for any  $x \in \mathbb{R}^d$ .

Consider the set of admissible trajectories

$$\mathcal{S}_{[0,\tau]}(x) := \{y_x \mid \dot{y}_x(s) \in F(y_x(s)) \text{ on } (0, \tau), y_x(0) = x\}$$

- ▶ Under **(H1)**,  $\mathcal{S}_{[0,\tau]}(x)$  is a compact subset of  $C([0, \tau])$ .
- ▶ Under **(H1)-(H2)**, the set-valued function  $x \rightsquigarrow \mathcal{S}_{[0,\tau]}(x)$  is Lipschitz continuous,

$$\exists L > 0, \mathcal{S}_{[0,\tau]}(x) \subset \mathcal{S}_{[0,\tau]}(z) + L|x - z|\mathcal{B}_{W^{1,1}} \quad \forall x, z \in \mathbb{R}^d.$$

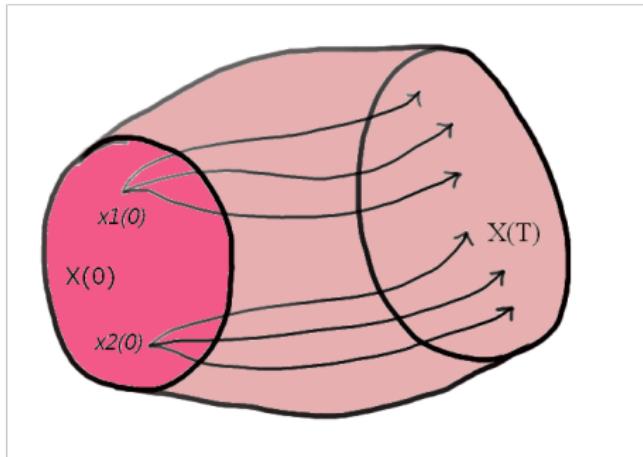
## Reachable (or Attainable) set

- The reachable set  $\mathcal{R}(x; t)$  from  $x$  at time  $t$  is the set of all points of the form  $y_x(\tau)$ , where  $y_x \in \mathcal{S}_{[0,t]}(x)$ :

$$\mathcal{R}(x; t) := \{y_x(\tau) \mid y_x \in \mathcal{S}_{[0,t]}(x), \tau \in [0, t]\}.$$

- The reachable set from  $X$  is defined by:

$$\mathcal{R}_X(t) := \cup_{x \in X} \mathcal{R}(x; t).$$



## Some properties of the reachable sets

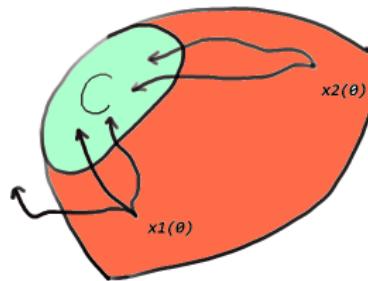
- If the set-valued  $F$  is usc with non-empty compact images (not necessarily convex images) then  $\mathcal{R}_C(t)$  is not necessarily closed.
- Assume (H1), then for every  $t \geq 0$ ,  $\mathcal{R}_C(t)$  is a closed set.

In all the sequel, we will assume that  $F$  satisfies at least (H1).

Let  $\mathcal{C}$  be a closed target set (in our examples,  $\mathcal{C}$  is **safe**)

## Capture Basin (or Backward reachable set)

- The Capture Basin  $C_{\sharp t}^{\mathcal{C}}$ , at time  $t$ , is the set of all initial positions  $x$  from which a trajectory  $y_x \in \mathcal{S}_{[0,t]}(x)$  can reach the target  $\mathcal{C}$ .

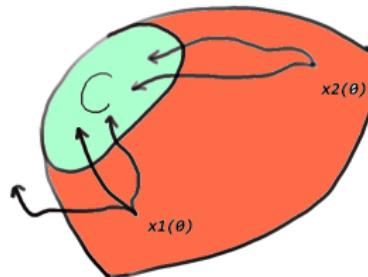


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## Capture Basin (or Backward reachable set)

- The Capture Basin  $\mathcal{R}_t^{\mathcal{C}}$ , **before time  $t$** , is the set of all initial positions  $x$  from which a trajectory  $y_x \in \mathcal{S}_{[0,t]}(x)$  can reach the target  $\mathcal{C}$  **before  $t$** .

$$\mathcal{R}_t^{\mathcal{C}} := \left\{ x \in \mathbb{R}^d, \exists \tau \in [0, t], \exists y_x \in \mathcal{S}_{[0,\tau]}(x), y_x(\tau) \in \mathcal{C} \right\}$$



# Optimization-based controller design

- Minimum-time control problem:

$$\mathcal{T}(x) = \inf \{ t; \quad y_x(t) \in \mathcal{C}, \quad y_x \in \mathcal{S}_{[0,t]}(x) \}.$$

# Optimization-based controller design

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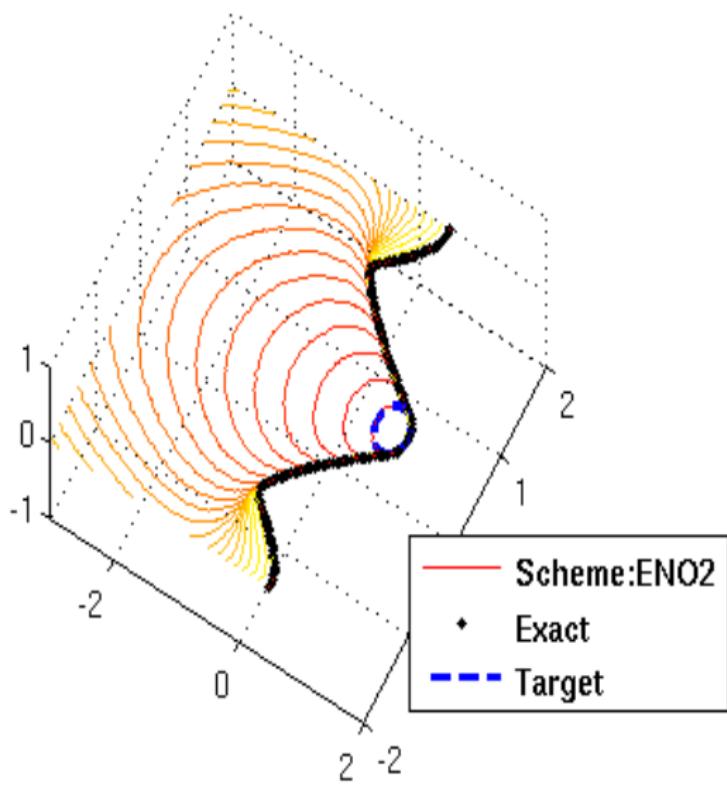
$$\mathcal{T}(x) = \inf \{ t; \quad y_x(t) \in \mathcal{C}, \quad y_x \in \mathcal{S}_{[0,t]}(x) \}.$$

- The sublevels of the minimum function  $\mathcal{T}$  correspond to the Capture Basins of the target  $\mathcal{C}$ :

$$C_{\sharp t}^{\mathcal{C}} = \{x \in \mathbb{R}^d \mid \mathcal{T}(x) = t\} \quad (3)$$

$$\mathcal{R}_t^{\mathcal{C}} = \{x \in \mathbb{R}^d \mid \mathcal{T}(x) \leq t\}. \quad (4)$$

# Zermelo problem



## Distance function. Level-set approach

Consider the function  $\Phi(x) = d_C(x)$ .

- A Mayer's problem:

$$V(x, t) = \inf_{y_x \in \mathcal{S}_{[0, t]}(x)} \Phi(y_x(t))$$

- A Mayer's problem with extended trajectories

$$\hat{V}(x, t) = \inf_{y_x \in \hat{\mathcal{S}}_{[0, t]}(x)} \Phi(y_x(t))$$

where  $\hat{\mathcal{S}}_{[0, t]}(x)$  is the set of trajectories satisfying:

$$\dot{y} \in \hat{F}(y(t)), \quad y(0) = x,$$

with  $\hat{F}(x) = [0, 1] \cdot F(x)$ .

## Level set approach

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## Level set approach

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- The minimum time function  $\mathcal{T} : \mathbb{R}^d \rightarrow \mathbb{R}^+ \cup \{+\infty\}$  is lsc.  
Moreover, we have:

$$\mathcal{T}(x) = \inf\{t \geq 0; V(x, t) \leq 0\} = \min\{t \geq 0; \hat{V}(x, t) \leq 0\}.$$

## Level set approach

- $V$  and  $\hat{V}$  are Lipschitz continuous functions
- For every  $t \geq 0$ ,

$$\mathcal{R}_{\sharp t}^{\mathcal{C}} = \{x \in \mathbb{R}^d; V(x, t) \leq 0\};$$

$$\mathcal{R}_t^{\mathcal{C}} = \{x \in \mathbb{R}^d; \hat{V}(x, t) \leq 0\}$$

- The minimum time function  $\mathcal{T} : \mathbb{R}^d \rightarrow \mathbb{R}^+ \cup \{+\infty\}$  is lsc.  
Moreover, we have:

$$\mathcal{T}(x) = \inf\{t \geq 0; V(x, t) \leq 0\} = \min\{t \geq 0; \hat{V}(x, t) \leq 0\}.$$

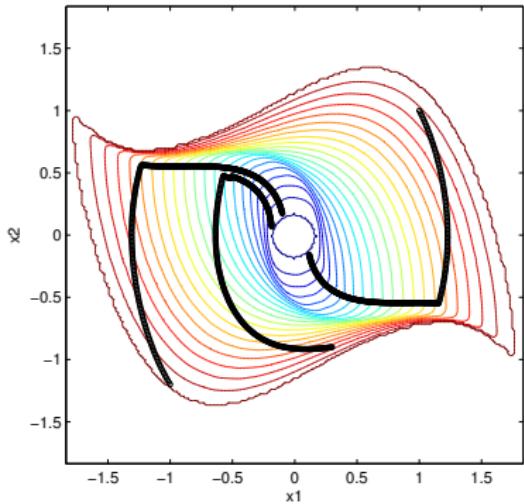
- $\Phi$  can be **any** function satisfying

$$\Phi(x) \leq 0 \iff x \in \mathcal{C}.$$

## Van der Pol Problem :

$$\begin{cases} \dot{y}_1(t) = y_2 \\ \dot{y}_2(t) = -y_1 + y_2(1 - y_1^2) + a(t) \\ a(t) \in [-1, 1] \end{cases}$$

$$\Phi(y) = d_{\mathcal{C}}(x)$$



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Now, consider the following control problems:

- Mayer's problem:

$$V(x, t) = \inf_{y_x \in \mathcal{S}_{[0,t]}(x)} \Phi(y_x(t))$$

- Time minimum problem ( $\mathcal{C}$  closed set in  $\mathbb{R}^d$ ):

$$\mathcal{T}(x) = \inf \{t; \quad y_x(t) \in \mathcal{C}, \quad y_x \in \mathcal{S}_{[0,t]}(x)\}$$

- Supremum cost:

$$V^\infty(x, t) = \inf_{y_x \in \mathcal{S}_{[0,t]}(x)} \Phi(y_x(t)) \bigvee \sup_{\theta \in [0, t]} g(y_x(\theta))$$

Assume (H1)-(H2) hold, and  $g : \mathbb{R}^d \rightarrow \mathbb{R}$  is Lipschitz continuous.

- If  $\phi : \mathbb{R}^d \rightarrow \mathbb{R}$  is lsc (resp. Lipschitz), then  $V$  and  $V^\infty$  are lsc (resp. Lipschitz).
- When the target  $\mathcal{C}$  is closed, the minimum time function  $\mathcal{T}$  is lsc.

## Mayer Problem

$$\begin{aligned} V(x, t) &= \min_{y_x \in \mathcal{S}_{[0, h]}(x)} V(y_x(h), t - h) \quad h \in (0, t), \\ V(x, 0) &= \Phi(x) \end{aligned}$$

## Mayer Problem

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$$V(x, 0) = \Phi(x)$$

## Minimum time problem:

$$\mathcal{T}(x) = \min_{y_x \in \mathcal{S}_{[0, h]}(x)} \mathcal{T}(y_x(h)) + h \quad h < \mathcal{T}(x), x \notin \mathcal{C},$$

$$\mathcal{T}(x) = 0 \quad x \in \mathcal{C};$$

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$$V^\infty(x, t) = \min_{y_x \in \mathcal{S}_{[0, h]}(x)} V^\infty(y_x(h), t - h) \bigvee \sup_{\theta \in [0, h]} g(y_x(\theta))$$

$$V^\infty(x, 0) = \Phi(x) \bigvee g(x);$$

$$V(x, t) = \min_{y_x \in \mathcal{S}_{[0,t]}(x)} \Phi(y_x(t))$$

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► **Suboptimality:**

$\forall y_x \in \mathcal{S}_{[0,t]}(x)$ ,  $s \longmapsto V(y_x(s), t - s)$  is increasing,

► **Superoptimality**

$\exists y_x^* \in \mathcal{S}_{[0,t]}(x)$ ,  $s \longmapsto V(y_x^*(s), t - s)$  is constant

Next, we derive the Hamilton-Jacobi-Bellman equation (HJB), which is an infinitesimal version of the DPP.

If the value function is  $C^1$  in a neighborhood of  $(x, t)$ , then

- $\partial_t V(x, t) + H(x, D_x V(x, t)) = 0, \quad x \in \mathbb{R}^d, \quad t > 0;$
- $H(x, D\mathcal{T}(x)) = 1, \quad x \notin \mathcal{C}, \quad \mathcal{T}(x) < +\infty;$
- $\min(\partial_t V^\infty(x, t) + H(x, DV^\infty(x, t)), V^\infty(x, t) - g(x)) = 0,$   
 $x \in \mathbb{R}^d, t > 0;$

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where  $H(x, q) := \max_{p \in F(x)} (-p \cdot q)$ .

Let  $u \in C_b^1(\mathbb{R}^d \times [0, T])$ . We say that  $u$  is a classical verification function if it satisfies

$$\begin{aligned}\partial_t u(x, t) + H(x, D_x u(x, t)) &\leq 0 \quad x \in \mathbb{R}^d, t \in [0, T], \\ u(x, 0) &= \Phi(x);\end{aligned}$$

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### Proposition

Let  $y_x^* \in S_{[0, T]}(x)$  be a feasible arc. Assume that there exists  $u$  a classical verification function, s.t.

$$u(x, T) = \Phi(y_x^*(T)),$$

then  $y_x^*$  is optimal for  $x$  on  $[0, T]$ .

## Proof.

► We take any  $y_x \in \mathcal{S}_{[0,T]}(x)$ , we have:

$$\frac{d}{ds}[u(y_x(s), T - s)] = -\partial_t u(y_x(s), T - s) + \dot{y}_x(s) \cdot Du(y_x(s), T - s) \geq 0.$$

Which means that  $s \mapsto u(y_x(s), T - s)$  is increasing:

$$u(x, T) \leq u(y_x(T), 0) = \Phi(y_x(T)).$$

Therefore,  $u(x, T) \leq V(x, T)$ .

►  $u(x, T) = \Phi(y_x^*(T)) \geq V(x, T)$ .

► Therefore  $u(x, T) = \Phi(y_x^*(T)) = V(x, T)$ , and  $y_x^*$  is optimal.

Assume  $F = f(x, A)$ ,  $A$  is a compact set in  $\mathbb{R}^m$ .

### Proposition

Let  $y_x^* \in S_{[0, T]}(x)$  be a feasible arc, and let  $a^*(\cdot) \in A$  be the corresponding control strategy:

$$\dot{y}_x^*(t) = f(y_x^*(t), a^*(t)) \quad \text{a.e in } [0, T].$$

Assume that there exists  $u$  a classical verification function, s.t.

$$\partial_t u(y_x^*(t), t) - f(y_x^*(t), a^*(t)).Du(y_x^*(t), t) = 0 \quad \text{for a.e } t \in [0, T],$$

then  $y_x^*$  is optimal for  $x$  on  $[0, T]$ .

# Proof.

- We know that  $s \mapsto u(y_x(s), T - s)$  is increasing:

$$u(x, T) \leq u(y_x(T), 0) = \Phi(y_x(T)).$$

Therefore,  $u(x, T) \leq V(x, T)$ .

- $u(y_x^*(s), T - s) = u(x, T) = Cst = u(y_x^*(T), 0) = \Phi(y_x^*(T)) \geq V(x, T)$ .
- Therefore  $u(x, T) = \Phi(y_x^*(T)) = V(x, T)$ , and  $y_x^*$  is optimal.

If we take the value function  $V$  itself as a verification function (i.e, if  $V$  is smooth), then a necessary and sufficient condition of optimality is:

$$\begin{aligned}-f(y_x^*(t), a^*(t)) \cdot DV(y_x^*(t), t) &= \max_{a \in A} (-f(y_x^*(t), a) \cdot DV(y_x^*(t), t)) \\ &= H(y_x^*(t), DV(y_x^*(t), t))\end{aligned}$$

for a.e  $0 < t < T$ .

Under the (irrealistic!) assumption that  $V$  is smooth, we consider the multivalued function:

$$\Psi(x, t) := \operatorname{argmax}_{a \in A} (-f(x, a) \cdot DV(x, t)).$$

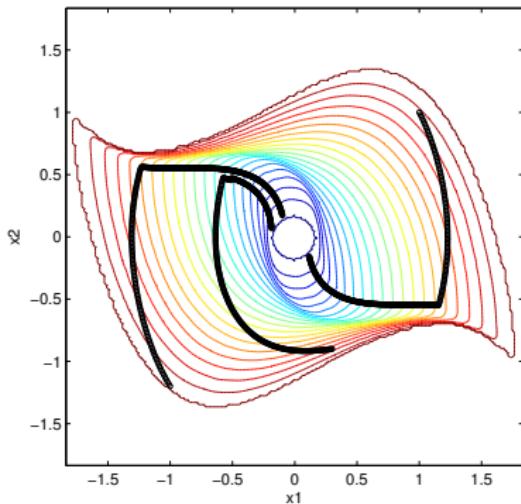
Let  $y_x^* \in \mathcal{S}_{[0, t]}(x)$ .  $y_x^*$  is optimal for  $x$  on  $[0, T]$  if and only if  $y_x^*(0) = x$ , and

$$\dot{y}_x^*(t) \in f(y_x^*(t), \Psi(y_x^*(t), t)) \quad t \in (0, T).$$

## Van der Pol Problem :

$$\begin{cases} \dot{y}_1(t) = y_2 \\ \dot{y}_2(t) = -y_1 + y_2(1 - y_1^2) + a(t) \\ a(t) \in [-1, 1] \end{cases}$$

$$\Phi(y) = \|y\| - r_0$$



The value function satisfies a Hamilton-Jacobi-Bellman equation (HJB) in a **viscosity sense**.

- $\partial_t V(x, t) + H(x, D_x V(x, t)) = 0, \quad x \in \mathbb{R}^d, \quad t > 0;$
- $H(x, D\mathcal{T}(x)) = 1, \quad x \notin \mathcal{C}, \quad \mathcal{T}(x) < +\infty;$
- $\min(\partial_t V^\infty(x, t) + H(x, DV^\infty(x, t)), V^\infty(x, t) - g(x)) = 0,$   
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where  $H(x, q) := \max_{p \in F(x)} (-p \cdot q)$ .

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where  $H(x, q) := \max_{p \in F(x)} (-p \cdot q)$ .

## Theorem

Assume that the value function is *continuous*. Then it is a viscosity solution of the corresponding HJB equation.

**Proof for the Minimum time problem.** Let  $\varphi \in C^1$  is such that  $\mathcal{T} - \varphi$  has a local maximum at  $x \notin \mathcal{C}$ , then for any  $y_x \in \mathcal{S}_{[0,s]}(x)$  we have:

$$\varphi(x) - \varphi(y_x(s)) \leq \mathcal{T}(x) - \mathcal{T}(y_x(s)) \leq s$$

for small  $s$ . Therefore,

$$\sup_{p \in F(x)} (-p \cdot D\varphi(x)) = H(x, D\varphi) \leq 1,$$

so  $\mathcal{T}$  is a subsolution. With similar arguments, we prove that  $\mathcal{T}$  is a supersolution as well.

- The boundedness assumption on  $V$  and  $V^\infty$  is not restrictive. It is verified whenever  $\Phi$  and  $g$  are bounded.
- The continuity of  $V$  and  $V^\infty$  holds whenever  $\Phi$  and  $g$  are continuous.
- The uniqueness results for the HJB equation associated to the minimal time problem is not a trivial task.

Assume  $\mathcal{C}$  is a closure of smooth domain, and let  $\eta_x$  be the normal to  $\mathcal{C}$ . If

$$\min_{p \in F(x)} p \cdot \eta_x < 0, \quad \forall x \in \partial\mathcal{C},$$

then the minimal time function  $\mathcal{T}$  is continuous.

The continuity of  $\mathcal{T}$  requires a *controllability* assumption of the system around the target. This important property is not satisfied in several examples, such as the Zermelo problem.

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$$\begin{cases} \partial_t V(x, t) + H(x, DV(x, t)) = 0 & x \in \mathbb{R}^d, t > 0 \\ V(x, 0) = \Phi(x) & x \in \mathbb{R}^d \end{cases}$$

where  $H(x, p) = \max_{q \in F(x)} (-q \cdot p)$ .

- A viscosity solution is given by the value function:

$$V(x, t) = \min_{y_x \in \mathcal{S}_{[0,t]}(x)} \Phi(y_x(t)).$$

- Regularity and growth properties of  $V$  can be derived directly from the definition of  $V$ . Moreover,  $V$  satisfies:
  - **Sub-optimality**:  $V(x, t) \leq V(y_x(h), t - h)$ ,  $\forall y_x \in \mathcal{S}_{[0,h]}(x)$ .
  - **Super-optimality**:  $\exists y_x^* \in \mathcal{S}_{[0,t]}(x)$ ,  $V(x, t) = V(y_x^*(h), t - h)$ .

# Comparison Principle: A geometric argument (nonsmooth analysis)

## Proposition

Let  $u : \mathbb{R}^d \times [0, T] \rightarrow \mathbb{R}$ . (i) Assume  $u$  is usc. Then  $u$  is a viscosity sub-solution iff  $u$  satisfies the sub-optimality principle.

(ii) Assume  $u$  is lsc. Then,  $u$  is a viscosity super-solution iff  $u$  satisfies the super-optimality principle.

## Idea of the proof

Let  $U : (0, T) \rightarrow \mathbb{R}$  a usc viscosity sub-solution of:

$$-U'(t) \leq 0 \quad t \in (0, T),$$

then  $U(t) \leq U(t - s)$  for every  $s \in [0, t]$ , for every  $0 \leq t \leq T$ .

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- For every  $p \in D^+ U(t)$ , we have  $-p \leq 0$
- Since  $u$  is usc:  $p \in D^+ U(t) \longleftrightarrow (p, 1) \in \mathcal{N}_{H^p(U)}(t, U(t))$ .

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- Since  $u$  is usc:  $p \in D^+ U(t) \longleftrightarrow (p, 1) \in \mathcal{N}_{H_p(U)}(t, U(t))$ .
- We conclude (by Rockafellar's Horizontal theorem) that:

$$(p, q) \cdot \begin{pmatrix} -1 \\ 0 \end{pmatrix} \leq 0, \quad \forall (p, q) \in \mathcal{N}_{H_p(U)}(t, U(t)). \quad (5)$$

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- For a given  $t > 0$ , consider the ODE:

$$\dot{Z}(s) = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \quad s \in (0, t), \quad Z(0) = \begin{pmatrix} t \\ U(t) \end{pmatrix}.$$

By (5) and the invariance principle, we conclude that

$$Z(s) \in H_p(U) \quad \forall s \in (0, t).$$

# Comparison principle: Doubling variable techniques (Kruzhkov's method for conservation laws)

## Theorem

Let  $\Omega$  be a bounded open set of  $\mathbb{R}^d$ . Let  $u_1, u_2 \in BUC(\bar{\Omega})$  be respectively, viscosity sub- and super-solutions of

$$u + H(x, Du(x)) = 0 \quad \text{on } \Omega,$$

and  $u_1 \leq u_2$  on  $\partial\Omega$ . Moreover, assume that  $H : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$  is uniformly continuous w.r.t the  $x$ -variable

$$|H(x_1, p) - H(x_2, p)| \leq \omega(|x_1 - x_2|(1 + |p|)),$$

where  $\omega : \mathbb{R}^+ \rightarrow \mathbb{R}^+$  is a continuous non-decreasing function satisfying  $\omega(0) = 0$ . Then  $u_1(x) \leq u_2(x)$  for every  $x \in \bar{\Omega}$ .

## Proof - 1: Case where $u_1, u_2$ are smooth ( $C^1$ functions)

- Assume  $u_1 - u_2$  attains a maximum at  $x_0 \in \Omega$ . Then  $p := Du_1(x_0) = Du_2(x_0)$  and:

$$u_1(x_0) + H(x_0, p) \leq 0 \leq u_2(x_0) + H(x_0, p).$$

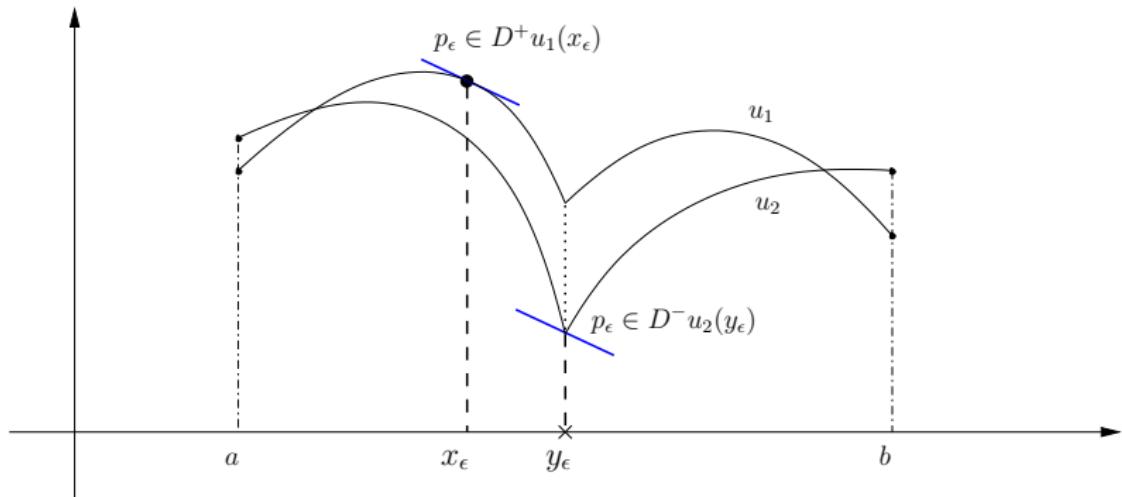
- We conclude that  $u_1(x_0) \leq u_2(x_0)$ .

## Proof - 2: the general case

Assume  $u_1 - u_2$  attains a positive maximum at  $x_0 \in \Omega$ .

- It may happen that  $D^+ u_1(x_0) \cap D^- u_2(x_0) = \emptyset$ . In this case, the HJB equation doesn't give any information at  $x_0$ .
- The **doubling variables** idea consists of building two sequences  $(x_\epsilon)$ ,  $(y_\epsilon)$  in a neighbourhood of  $x_0$ , such that

$$D^+ u_1(x_\epsilon) \cap D^- u_2(y_\epsilon) \neq \emptyset$$



$$\exists p_\epsilon \in D^+ u_1(x_\epsilon) \cap D^- u_2(y_\epsilon)$$

## Proof - 2: the general case

- Assume  $u_1 - u_2$  attains a positive maximum at  $x_0 \in \Omega$ . Set

$$\delta := u_1(x_0) - u_2(x_0) > 0.$$

## Proof - 2: the general case

- Assume  $u_1 - u_2$  attains a positive maximum at  $x_0 \in \Omega$ . Set

$$\delta := u_1(x_0) - u_2(x_0) > 0.$$

- Define the function of two variables:

$$\Psi_\varepsilon(x, y) := u_1(x) - u_2(y) - \frac{|x - y|^2}{2\varepsilon}.$$

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- $\Psi$  attains its global maximum  $(x_\varepsilon, y_\varepsilon)$  on  $\overline{\Omega} \times \overline{\Omega}$ , and we have:

$$\Psi(x_\varepsilon, y_\varepsilon) \geq \Psi(x_0, x_0) \implies \Psi(x_\varepsilon, y_\varepsilon) \geq \delta > 0. \quad (6)$$

► Let  $M > 0$  s.t  $\|u_1\|_\infty + \|u_2\|_\infty \leq M$ . We get:

$$0 < \Psi(x_\varepsilon, y_\varepsilon) \leq M - \frac{|x_\varepsilon - y_\varepsilon|^2}{2\varepsilon},$$

which implies that:  $|x_\varepsilon - y_\varepsilon| \leq \sqrt{2M\varepsilon}$ .

Moreover,  $\frac{x_\varepsilon - y_\varepsilon}{\varepsilon} \rightarrow 0$ , when  $\varepsilon \rightarrow 0$ .

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which implies that:  $|x_\varepsilon - y_\varepsilon| \leq \sqrt{2M\varepsilon}$ .

Moreover,  $\frac{x_\varepsilon - y_\varepsilon}{\varepsilon} \rightarrow 0$ , when  $\varepsilon \rightarrow 0$ .

► Since  $u_2 \in \text{BUC}(\overline{\Omega})$ , there exists  $\varepsilon' > 0$  s.t.:

$$|u_2(x) - u_2(y)| \leq \delta/2 \quad \text{whenever } |x - y| \leq \sqrt{2M\varepsilon'}.$$

For every  $\varepsilon < \varepsilon'$ , we have:  $x_\varepsilon, y_\varepsilon \notin \partial\Omega$ . Indeed, if  $x_\varepsilon \in \partial\Omega$ , then we would have:

$$\begin{aligned}\Psi(x_\varepsilon, y_\varepsilon) &\leq (u_1(x_\varepsilon) - u_2(x_\varepsilon)) + |u_2(x_\varepsilon) - u_2(y_\varepsilon)| - \frac{|x_\varepsilon - y_\varepsilon|^2}{2\varepsilon} \\ &\leq 0 + \delta/2 + 0\end{aligned}$$

in contradiction with (6).

► We can check that  $p_\varepsilon := \frac{x_\varepsilon - y_\varepsilon}{\varepsilon} \in D^+ u_1(x_\varepsilon) \cap D^- u_2(y_\varepsilon)$

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- From the definition of  $u_1, u_2$ , we obtain:

$$u_1(x_\varepsilon) + H(x_\varepsilon, p_\varepsilon) \leq 0 \leq u_2(y_\varepsilon) + H(y_\varepsilon, p_\varepsilon).$$

- We can check that  $p_\varepsilon := \frac{x_\varepsilon - y_\varepsilon}{\varepsilon} \in D^+ u_1(x_\varepsilon) \cap D^- u_2(y_\varepsilon)$
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$$u_1(x_\varepsilon) + H(x_\varepsilon, p_\varepsilon) \leq 0 \leq u_2(y_\varepsilon) + H(y_\varepsilon, p_\varepsilon).$$

- Finally, by using the assumption on  $H$ , it comes:

$$\delta \leq \Psi(x_\varepsilon, y_\varepsilon) \leq u_1(x_\varepsilon) - u_2(y_\varepsilon) \leq \omega(|x_\varepsilon - y_\varepsilon|(1 + |p_\varepsilon|)).$$

This yields a contradiction when  $\varepsilon$  goes to 0.

# Time dependent case.

## Theorem

Let  $H : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$  is Lipschitz continuous:

$$\begin{aligned}|H(x, p) - H(y, p)| &\leq C(|x - y|(1 + |p|)), \\ |H(x, p_1) - H(x, p_2)| &\leq C|p_1 - p_2|.\end{aligned}$$

Let  $u_1, u_2$  are resp., a sub- and super-solution of

$$\partial_t u + H(x, Du) = 0 \quad x \in \mathbb{R}^d, \quad t \in (0, T]$$

Assume that  $u_1(x, 0) \leq u_2(x, 0)$ , then  $u_1 \leq u_2$  on  $\mathbb{R}^d \times [0, T]$ .

## Idea of the proof.

- A new difficulty: the domain is unbounded. For this, we define the *doubling-variable* function as:

$$\Psi_\varepsilon(x, t, y, s) := u_1(x, t) - u_2(y, s) - \varepsilon(|x|^2 + |y|^2) - \frac{(|t - s|^2 + |x - y|^2)}{2\varepsilon^2}.$$

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$$\begin{cases} \partial_t V(x, t) + H(x, DV(x, t), D^2 V(x, t)) = 0 & x \in \mathbb{R}^d, t > 0 \\ V(x, 0) = \Phi(x) & x \in \mathbb{R}^d \end{cases}$$

where

$$H(x, p, Q) = \max_{a \in A} (-\ell(x, a) - b(x, a) \cdot p - \frac{1}{2} \text{Tr}(\sigma(x, a) \sigma^T(x, a) \cdot Q)).$$

$\sigma(\cdot)$  is a mapping  $\mathbb{R}^d \times A$  into the space of  $n \times r$  matrices,  $b : \mathbb{R}^d \times A \rightarrow \mathbb{R}^d$ .

## Definition.

Let  $\Omega \subset \mathbb{R}^m$ , and  $\mathcal{F} : \Omega \times \mathbb{R} \times \mathbb{R}^m \times \mathcal{S}_{m \times m} \rightarrow \mathbb{R}$  be continuous and satisfying  $\mathcal{F}(x, r, p, X) \leq \mathcal{F}(x, r, p, Y)$  when  $X \geq Y$ .

Consider the HJB equation:

$$\mathcal{F}(x, u(x), Du(x), D^2u(x)) = 0 \quad \text{in } \Omega. \quad (7)$$

(i) A function  $u \in USC(\Omega)$  is a viscosity **sub-solution** of (7) if for any  $\varphi \in C^2(\Omega)$  and any local maximum point  $x_0 \in \Omega$  of  $u - \varphi$ ,

$$\mathcal{F}(x_0, u(x_0), D\varphi(x_0), D^2\varphi(x_0)) \leq 0.$$

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(iii) A function  $u \in C(\Omega)$  is a viscosity **solution** of (7) if it is a sub- and super-solution.

- The notion of viscosity can also be defined by an adequate concept of "super-" and "sub-" derivatives (of order 2), called "semi-jets".
- The viscosity solution is the limit of classical solutions of *viscous* equations.
- As for first order equations, a general stability property holds, without any requirement about the convergence of derivatives.

## Time-dependent HJ equation

- The associated control problem:

$$\begin{aligned} V(x, t) &= \text{Min } \mathbb{E} \left[ \Phi(y(T)) + \int_0^t \ell(y(s), \alpha(s)) ds \right]; \\ dy(s) &= b(y(s), \alpha(s))ds + \sigma(y(s), \alpha(s))dw_s, \\ y(0) &= x, \\ \alpha(s) &\in A, \quad s \in [0, t]. \end{aligned}$$

- $V$  satisfies a DPP:

$$\begin{aligned} V(x, t) &= \min_{\alpha([0, h] \subset A)} \mathbb{E} \left[ V(y_x^\alpha(h), t-h) + \int_0^h \ell(y_x^\alpha(s), \alpha(s)) ds \right], \\ V(x, 0) &= \Phi(x). \end{aligned}$$

- A general comparison principle can be stated, as for first order equations, under Lipschitz continuity of the Hamiltonian.

# References on viscosity notion for HJ equations

-  G. Barles, *Solutions de Viscosité des Equations de Hamilton-Jacobi*, Springer, 1994.
-  M. Bardi and I. Capuzzo Dolcetta, *Optimal Control and Viscosity Solutions of Hamilton-Jacobi-Bellman Equations*, Birkhäuser, 1997.
-  M. Crandall and P. L. Lions, Viscosity solutions of Hamilton-Jacobi equations, *Trans. Amer. Math. Soc.*, 277(1983), 1-42.
-  M. Crandall, H. Ishii and P. L. Lions, User's guide to viscosity solutions of second order partial differential equations, *Bull. Amer. Math. Soc.*, 27(1992), 1-67.
-  W.H. Fleming and H.M. Soner, *Controlled Markov Processes and Viscosity Solutions*. Springer-Verlag, New-York, 1993.
-  S. N. Kruskov, First order quasilinear equations in several independant variables. *Math. USSR-Sbornik* 10(1970), 217-243.
-  P.L. Lions, *Generalized solution of Hamilton-Jacobi equations*, Pitman, London, 1982.

## Extensions and active fields

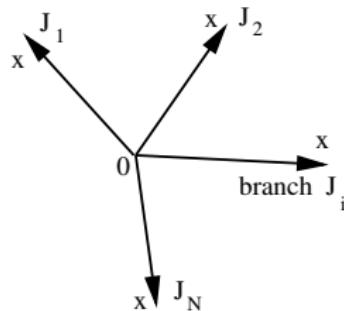
- The comparison principle has been extended to unbounded viscosity solutions, with polynomial or exponential growth (ref: Ishii'92)
- Boundary conditions can be also considered in the viscosity framework (ref: Barles, Capuzzo-Dolceta, Lions, Ishii, ... )
- Some existence and uniqueness results have been extended to Hamilton-Jacobi equations with  $L^p$  data (ref: Caffarelli, Cardaliaguet, Souganidis, ... )
- Hamilton-Jacobi equations in infinite dimension setting have been investigated by (ref: Barbu, Cannarsa, Gozzi, Soner, ... )
- Homogenization and Ergodic control problems (ref: Alvarez, Bardi, Achdou, Cardaliaguet, Siconolfi, Souganidis ... )
- ....

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# A junction condition on $O$

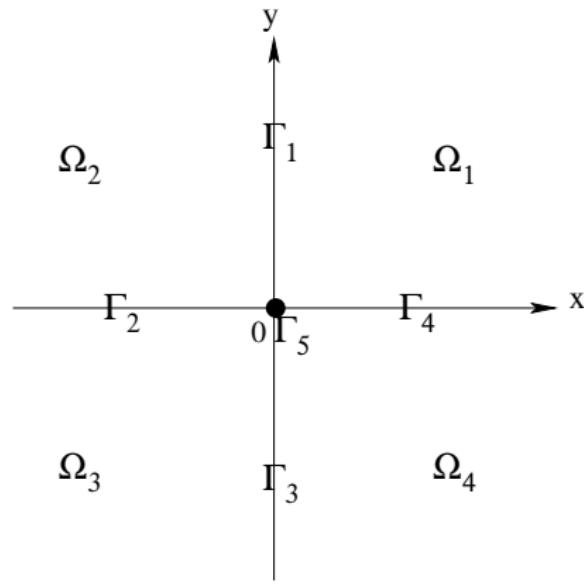


$$\left\{ \begin{array}{ll} u_t^i + H_i(u_x^i) = 0, & x > 0, \quad i = 1, \dots, N \\ u^i = u^j =: u, & x = 0, \\ u_t + F(u_x^1, \dots, u_x^N) = 0, & x = 0 \end{array} \right.$$

## Junction conditions on multi-domains

- $\mathbb{R}^d$  is divided into several open disjoint subdomains  $(\Omega_i)_{i=1,\dots,m}$  with

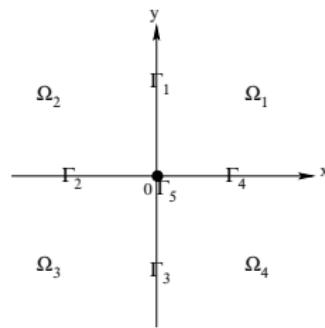
$$\mathbb{R}^d = \overline{\Omega}_1 \cup \cdots \cup \overline{\Omega}_m.$$



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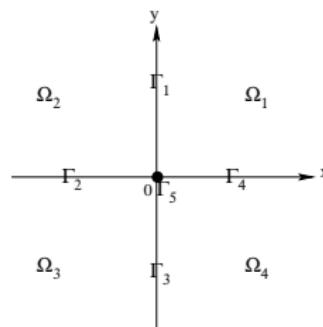


$$\begin{cases} \partial_t u_i(t, x) + H_i(x, D_x u_i(t, x)) = 0 & t \in (0, T), x \in \Omega_i, \\ u_i(0, x) = \varphi(x) & x \in \mathbb{R}^d, \end{cases}$$

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$$u = u_1, \dots, u_m$$

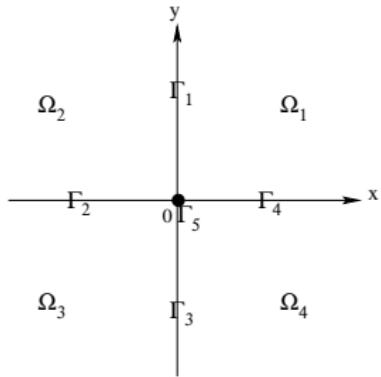
$$\partial_t u(t, x) + F(x, D_x u_1(t, x), \dots, D_x u_m(t, x)) = 0 \quad \text{on } \cup_j \Gamma_j$$

# Motivation & Position of the problem

- data transmission and traffic management  
(Garavello-Piccoli'2006, Engel et al.'08, Lebacque'05)
- blood circulation
- supply chains (Göttlich , Herty , Klar 2005)
- gas pipelines
  
- multiprocess systems without transition cost
- multiple thermostatic hybrid systems (Bagagiolo et al.'2012)
- Numerical purpose: decomposition of domains

# Position of the problem

$$\triangleright \mathbb{R}^d = \left( \bigcup_{i=1}^m \Omega_i \right) \cup \left( \bigcup_{j=1}^{\ell} \Gamma_j \right),$$



➤ For each  $i = 1, \dots, m$ , consider the system of HJB equations:

$$\begin{cases} \partial_t u_i(t, x) + H_i(x, D_x u_i(t, x)) = 0 & t \in (0, T), x \in \Omega_i, \\ u(0, x) = \varphi(x) & x \in \mathbb{R}^d, \end{cases} \quad (8)$$

where  $T > 0$  is a fixed final time and  $\varphi$  is a given regular function.

➤ Each  $H_i$  is a **Lipschitz continuous** Hamiltonian of Bellman form:

$$H_i(x, p) := \sup_{v \in F_i(x)} \{-v \cdot p - L_i(x, v)\}.$$

where  $F_i$  is a Lipschitz set-valued function on  $\Omega_i$ .

## Junction conditions

**Question:** what conditions of transition between the subdomains  $\Omega_i$  should be considered to ensure the existence and *uniqueness* of solution  $u$  for the system (8)?

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For simplicity, we shall take  $L \equiv 0$ .

## Junction conditions

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Consider the equation

$$\begin{aligned}\partial_t u(t, x) + H_F(x, D_x u(t, x)) &= 0, \quad x \in \Omega = \mathbb{R}^d \\ u(0, x) &= \varphi(x),\end{aligned}$$

where  $H_F(x, q) = \sup_{p \in F(x)} \{-p \cdot q\}$ .

### Definition: viscosity solution

A continuous function  $u$  is subsolution if for every  $\Phi \in C^1((0, T) \times \Omega)$  such that

$u \leq \Phi$  (resp.  $u \geq \Phi$ ) with equality at  $(t_o, x_o) \in \Omega$ , then:

$$\partial_t \Phi(t_o, x_o) + H_F(x_o, D_x \Phi(t_o, x_o)) \leq 0 \quad (\text{resp. } \geq 0).$$

# Mayer's optimal control problems

- Assume:  $\mathbb{R}^d = \Omega$
- Let  $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$  be a regular function and  $F : \mathbb{R}^d \rightsquigarrow \mathbb{R}^d$  be a **Lipschitz continuous** multi-function with closed convex images. For an initial data  $(t, x) \in [0, T] \times \mathbb{R}^d$ , consider the differential inclusion:

$$\begin{cases} \dot{y}(s) \in F(y(s)) & s \in (0, t), \\ y(0) = x. \end{cases} \quad (9)$$

The set  $S_{[0,t]}(x)$  of all trajectories satisfying (9) is closed in  $W^{1,1}(0, t)$ .

- And the **value function**  $v : [0, T] \times \mathbb{R}^d \rightarrow \mathbb{R}$  is defined by

$$v(t, x) := \inf\{\varphi(y(t)) : y(\cdot) \text{ satisfies (9)}\}.$$

# Classical Dynamical Programming Principle

For any  $t \in [0, T]$ ,  $x \in \mathbb{R}^d$ ,  $h \in [0, t]$ , we have

$$v(t, x) = \min_{y \in \mathcal{S}_{[0,t]}(x)} \{v(t - h, y(h))\}.$$

More precisely,

- **The super-optimality:** There exists an optimal trajectory  $\bar{y} \in \mathcal{S}_{[0,t]}(x)$  such that

$$v(t, x) \geq (=) v(t - h, \bar{y}(h));$$

- **The sub-optimality:** For any trajectory  $y \in \mathcal{S}_{[0,t]}(x)$ , we have:

$$v(t, x) \leq v(t - h, y(h)).$$

# Characterization

In the classical case where  $F$  is Lipschitz continuous, consider the Hamiltonian

$$H_F(x, p) = \sup_{q \in F(x)} \{-q \cdot p\}.$$

For any Lipschitz continuous function  $u : [0, T] \times \mathbb{R}^d \rightarrow \mathbb{R}$ , we have

- ▶  $u$  satisfies the super-optimality  $\Leftrightarrow u$  satisfies

$$\partial_t u(t, x) + H_F(x, Du(t, x)) \geq 0;$$

- ▶  $u$  satisfies the sub-optimality  $\Leftrightarrow u$  satisfies

$$\partial_t u(t, x) + H_F(x, Du(t, x)) \leq 0.$$

## Classical case

The value function  $v$  defined by:

$$v(t, x) := \min_{y \in S_{[0,t]}(x)} \varphi(y(t)),$$

is the unique viscosity solution of

$$\begin{aligned}\partial_t v(t, x) + H_F(x, Dv(t, x)) &= 0 \\ v(0, x) &= \varphi(x).\end{aligned}$$

# Consider the cross structure in $\mathbb{R}^2$

$$\begin{aligned}\partial_t u(t, x) + H_i(x, Du(t, x)) &= 0 \quad (0, T) \times \Omega_i \\ u(0, x) &= \varphi(x),\end{aligned}$$

with  $H_i(x, p) = \sup_{v \in F_i(x)} (-v \cdot p)$ .

- ▶ Idea: introduce a global optimal control problem and investigate the equation satisfied by the value function on the interfaces.

Main difficulty:

- ▶ How to define the dynamics on the interfaces which allow that the trajectories can either transit between the domains or stay on the interfaces?
- ▶ A global Lipschitz continuous Hamiltonian could not be expected due to the singularity of the domain.

# Literature on non-Lipschitz Hamiltonians

- ▶ Hamilton-Jacobi equations with discontinuous Hamiltonians: Ishii'89.
- ▶ Hamilton-Jacobi-Bellman equations with discontinuous Lagrangians: Soravia'02.
- ▶ Hamilton-Jacobi equations with measurable Hamiltonians: Camilli-Siconolfi'03.
- ▶ Infinite horizon optimal control problem on two-domains: Barles-Briani-Chasseigne'12.
- ▶ Hamilton-Jacobi approach to junction problems on network: Achdou-Camilli-Cutri-Tchou'12, Imbert-Monneau-Zidani'12.

# Assumptions

We are interested in the Hamiltonians  $H_i$  of the following [Bellman](#) form:

$$H_i(x, q) = \sup_{p \in F_i(x)} \{-p \cdot q\},$$

where  $F_i : \bar{\Omega}_i \rightsquigarrow \mathbb{R}^2$  are multifunctions satisfying the following: for any  $x \in \bar{\Omega}_i$ ,

- ▶  $F_i(x)$  is a nonempty, convex and compact set;
- ▶  $F_i$  is Lipschitz continuous with respect to Hausdorff metric;
- ▶ A controllability assumption. For example:  $\exists \delta > 0$  so that  $B(0, \delta) \subset F_i(x)$ .

## Filippov regularization for the dynamics

To define the global dynamics, we consider the following Filippov regularization of the multi-functions  $(F_i)_{i=1,\dots,4}$ :

$$FC(x) := \begin{cases} F_i(x) & \text{if } x \in \Omega_i, \\ \overline{\text{co}}(F_i(x) : x \in \overline{\Omega}_i) & \text{otherwise.} \end{cases}$$

Then consider the differential inclusion

$$\begin{cases} \dot{y}(s) \in FC(y(s)) & s \in (0, t), \\ y(0) = x. \end{cases} \quad (10)$$

**Remark:**  $FC$  is the smallest upper semi-continuous envelope of  $(F_i)_{i=1,\dots,4}$ . Moreover,

$$FC(x) = \bigcap_{\varepsilon > 0} \overline{\text{co}} \left\{ F_i(y) : \|y - x\| < \varepsilon, y \in \overline{\Omega}_i \right\}.$$

# Global optimal control problem

For any  $t \in [0, T]$ ,  $x \in \mathbb{R}^2$ , we set

$$S_{[0,t]}(x) := \{y(\cdot) | y(t) = x, \dot{y}(s) \in FC(y(s)) \text{ for } s \in (0, t)\}$$

as the set of trajectories. Consider the value function

$$v(t, x) := \inf_{y(\cdot) \in S_{[0,t]}(x)} \{\varphi(y(t))\}.$$

The upper semi-continuity and the convexity of  $FC$  imply that  $S_{[0,t]}(x)$  is compact, then the "inf" is in fact "min".

**Remark:** Here,  $v$  is the value function of a control problem of hybrid systems without transition cost  $\Rightarrow$  Zeno effect

# Dynamical Programming Principle

For any  $t \in [0, T]$ ,  $x \in \mathbb{R}^2$ ,  $h \in [0, t]$ , we have

$$v(t, x) = \min_{y(\cdot) \in S_{[t, T]}(x)} \{v(t - h, y(h))\}.$$

More precisely,

- **The super-optimality:**  $\exists \bar{y} \in S_{[t, T]}(x)$  such that

$$v(t, x) \geq (=) v(t - h, \bar{y}(h));$$

- **The sub-optimality:**  $\forall y \in S_{[t, T]}(x)$  such that

$$v(t, x) \leq v(t - h, y(h)).$$

## Problem of characterization

$FC$  is only upper semi-continuous, then the characterization of the **sub-optimality** fails, i.e.  $v$  does not satisfy

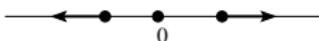
$$\partial_t v(t, x) + \sup_{p \in FC(x)} \{-p \cdot Dv(t, x)\} \leq 0.$$

From a point view of trajectories: the set of dynamics  $FC(x)$  is too large and may contain some useless dynamics which are never used by the trajectories.

Examples in dimension 1 with one interface  $\{0\}$ :

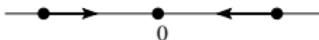
- ▶ The case "pull-pull":

$F_1(x) = [-1, \frac{1}{2}]$  for  $x < 0$ ,  $F_2(x) = [-\frac{1}{2}, 1]$  for  $x > 0$ . The convexification on 0 is  $FC(0) = [-1, 1]$ .



- ▶ The case "push-push":

$F_1(x) = \{1\}$  for  $x < 0$ ,  $F_2(x) = \{-1\}$  for  $x > 0$ . The convexification on 0 is also  $FC(0) = [-1, 1]$  which is not the reasonable set of dynamics.



We need to consider only the "useful" dynamics!

# Essential dynamics

Some notations:

- $\mathcal{T}_{\overline{\Omega}_i}(x)$  is the tangent cone of  $\overline{\Omega}_i$  on  $x$ ,
- $\mathcal{T}_\Gamma$  the tangent space of  $\Gamma$  on  $x$ .

Introduce the **essential dynamic** multifunction  $F^E : \mathbb{R}^d \rightsquigarrow \mathbb{R}^d$  (ref. Barnard-Wolenski'13, Rao-HZ'13):

- ▶ For  $x \in \Omega_i$ ,  $F^E(x) := F_i(x)$ ;
- ▶ For  $x \in \Gamma_j$ ,

$$F^E(x) = (FC(x) \cap \mathcal{T}_\Gamma(x)) \bigcup \bigcup_{i|x \in \overline{\Omega}_i} (F_i(x) \cap \mathcal{T}_{\overline{\Omega}_i}(x)).$$

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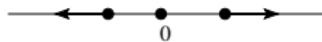
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$$F^E(x) = (FC(x) \cap \mathcal{T}_{\Gamma_j}(x)) \bigcup \bigcup_{i|x \in \overline{\Omega}_i} (F_i(x) \cap \mathcal{T}_{\overline{\Omega}_i}(x)).$$

- $F_i(x) \cap \mathcal{T}_{\overline{\Omega}_i}(x)$ : the dynamics in  $F_i(x)$  are inward to  $\overline{\Omega}_i$ ,
- $F(x) \cap \mathcal{T}_{\Gamma_j}(x)$ : the dynamics in  $F(x)$  are tangent to  $\Gamma_j$ .

# Examples in dimension 1

- ▶ Example of "pull-pull":

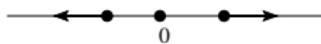


$$F_1(0) \cap \mathcal{T}_{\bar{\Omega}_1}(0) = \{-1\}, \quad F_2(0) \cap \mathcal{T}_{\bar{\Omega}_2}(0) = \{1\}, \quad F(0) \cap \mathcal{T}_F(0) = \{0\},$$

$$F^E(0) = \{-1, 0, 1\}.$$

# Examples in dimension 1

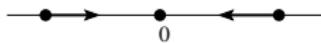
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- ▶ Example of "push-push":



$$F_1(0) \cap \mathcal{T}_{\bar{\Omega}_1}(0) = F_2(0) \cap \mathcal{T}_{\bar{\Omega}_2}(0) = \emptyset, \quad F(0) \cap \mathcal{T}_\Gamma(0) = \{0\},$$

$$F^E(0) = \{0\}.$$

## Examples in dimension 2

$\Omega_1 = \{(x, y) | x < 0, y \in \mathbb{R}\}$ ,  $\Omega_2 = \{(x, y) | x > 0, y \in \mathbb{R}\}$ ,  
 $\Gamma = \{(0, y) | y \in \mathbb{R}\}$ .

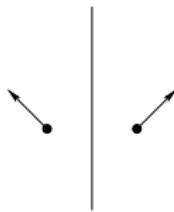


Figure : \*

Original dynamics:

$$F_1 = \{(-1, 1)\}, \\ F_2 = \{(1, 1)\}.$$

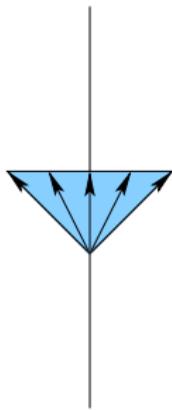


Figure : \*

Convexified dynamics

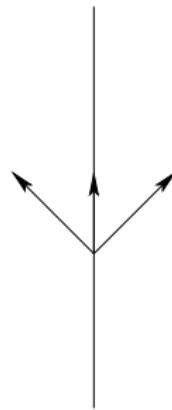


Figure : \*

Essential dynamics

## Properties of $F^E$

- $F^E(x) \subset FC(x)$ .
- The set of trajectories:  
 $\{y \mid \dot{y}(s) \in FC(y(s)); y(t) = x\} = \{y \mid \dot{y}(s) \in F^E(y(s)); y(t) = x\}$ .
- For any  $p \in F^E(x)$ , there exists  $y \in S_{[t;T]}(x)$  such that  $y(t) = x$  and  $\dot{y}(t) = p$ .

# Equations on the interface

Consider the new Essential Hamiltonian

$$H^E(x, q) := \sup_{p \in F^E(x)} \{-p \cdot q\}.$$

And here is the existence result:

## Theorem (Rao-HZ'12)

*The value function  $v$  is Lipschitz continuous and is the viscosity solution of the system of HJB equations:*

$$\begin{cases} \partial_t u(t, x) + H_i(x, Du(t, x)) = 0 & t \in (0, T), x \in \Omega_i, \\ u(0, x) = \varphi(x) & x \in \mathbb{R}^2. \end{cases}$$

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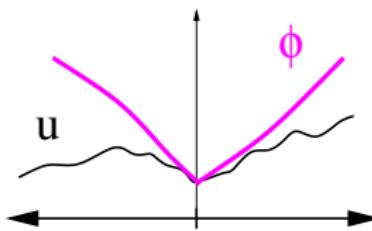
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Lipschitz continuity comes from the controllability assumption!



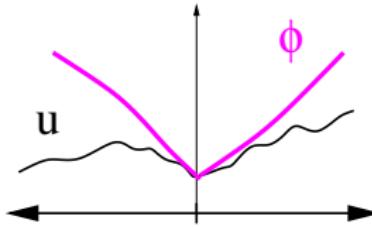
## Definition: viscosity solution (for $d = 1$ )

A continuous function  $u$  is subsolution if for every  $\Phi \in C^0((0, T) \times \mathbb{R}^d)$  such that

$\Phi \in C^1((0, T) \times \mathcal{M})$ , for  $\mathcal{M}^r = \mathbb{R}^+$ ,  $\mathcal{M}^l = \mathbb{R}^-$  or  $\mathcal{M}^O = \{0\}$ ,

$u \leq \Phi$  (resp.  $u \geq \Phi$ ) with equality at  $(t_o, x_o) \in \Omega$ , then:

$$\partial_t \Phi(t_o, x_o) + \sup_{k=l,r,O} H_k(x_o, D_x^k \Phi(t_o, x_o)) \leq 0 \quad (\text{resp. } \geq 0),$$



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where  $\mathbb{I}(x_o) := \{i \mid x \in \overline{\Omega}_i\} \cup \{j \mid x \in \Gamma_j\}$ .

## Super-solution: Idea of the proof (I)

The value function  $v$  is viscosity **super-solution** of:

$$\begin{cases} \partial_t u(t, x) + H^E(x, Du(t, x)) \geq 0 & t \in (0, T), x \in \mathbb{R}^d, \\ u(0, x) \geq \varphi(x) & x \in \mathbb{R}^d. \end{cases}$$

- for every  $(t, x) \in (0, T) \times \mathbb{R}^d$ , there exists  $\bar{y}_x \in S_{[0,t]}(x)$ :

$$v(t, x) \geq v(t - h, \bar{y}(h)) \quad \forall h \in (0, t).$$

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- if  $v \geq \Phi$  with equality at  $(t_o, x_o)$ , then

$$\Phi(t_o, x_o) \geq \Phi(t_o - h, \bar{y}_{x_o}(h)) \quad \forall h \in (0, t_o).$$

## Super-solution: Idea of the proof (II)

- On the other hand, for any  $x \in \mathbb{R}^d$ , and every  $y \in S_{[0,t]}(x)$ , we have (whenever the limit exists):

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$$\partial_t \Phi(t_o, x_o) + \sup_{p \in F^E(x_o)} (p \cdot D\Phi(t_o, x_o)) \geq 0.$$

- By separation Theorem, we get:

$$\partial_t \Phi(t_o, x_o) + H^E(x_o, D\Phi(t_o, x_o)) \geq 0.$$

## Sub-solution:Idea of the proof

### Theorem (Rao-Siconolfi-HZ'13)

*The value function  $v$  is viscosity **sub-solution** of the system of HJB equations:*

$$\begin{cases} \partial_t u(t, x) + H^E(x, Du(t, x)) \leq 0 & t \in (0, T), x \in \mathbb{R}^d, \\ u(0, x) \leq \varphi(x) & x \in \mathbb{R}^d. \end{cases}$$

- for every  $(t, x) \in (0, T) \times \mathbb{R}^d$ , for every  $\bar{y}_x \in \mathcal{S}_{[0,t]}(x)$ :

$$v(t, x) \leq v(t - h, \bar{y}(h)) \quad \forall h \in (0, t).$$

- if  $v \leq \Phi$  with equality at  $(t_o, x_o)$ , then

$$\Phi(t_o, x_o) \leq \Phi(t_o - h, \bar{y}_{x_o}(h)) \quad \forall h \in (0, t_o).$$

- Therefore

$$\partial_t \Phi(t_o, x_o) + H^E(x_o, D\Phi(t_o, x_o)) \leq 0.$$

# The uniqueness result

A strong comparison principle:

## Theorem

Let  $u_1$  and  $u_2$  be respectively a Lipschitz continuous sub-solution and super-solution of (2), then we have

$$u_1(t, x) \leq u_2(t, x), \text{ for } t \in [0, T], x \in \mathbb{R}^d.$$

Main steps of the proof:

- ▶ The subsolution  $u_1$  satisfies the sub-optimality  $\Rightarrow u_1 \leq v$ ;
- ▶ The supersolution  $u_2$  satisfies the super-optimality  $\Rightarrow u_2 \geq v$ ;

## First step: proof for the characterization of super-optimality

Since  $F^E(x) \subset FC(x)$ , the super-solution inequation

$$\partial_t u(t, x) + \sup_{p \in F^E(x)} (p \cdot Du(t, x)) \geq 0$$

implies that

$$\partial_t u(t, x) + \sup_{p \in FC(x)} (p \cdot Du(t, x)) \geq 0$$

Then the following of the proof is classical since  $FC$  is upper semi-continuous.

## Second step: The characterization of the sub-optimality is more technical ...

- For a trajectory starting in  $\Omega_i$  and living in the same manifold : NO Problem

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  - take a sliding mode and follow the same interface for a while: Junction conditions include the behavior of the trajectories on each interface

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- For a trajectory starting in  $\Omega_i$  and living in the same manifold : NO Problem
- When a trajectory reaches an interface, it can
  - cross the interface and pass to another manifold: continuity of the function
  - take a sliding mode and follow the same interface for a while: Junction conditions include the behavior of the trajectories on each interface
- The set of trajectories contains also some trajectories with more complicated behavior: Crucial difficulty

## An explicit example in 1d.

The multi-domains:  $\Omega_1 = \{x : x < 0\}$ ,  $\Omega_2 = \{x : x > 0\}$ ,  $\Gamma = \{0\}$ .

The dynamics:  $F_1 = [-\frac{1}{2}, 1]$ ,  $F_2 = [-1, \frac{1}{2}]$ .

$T > 0$  is a given final time, the final cost function  $\varphi(x) = x$ .

$$v_1(t, x) := \min\{y_{t,x}(T)\} = \begin{cases} x - \frac{1}{2}(T-t) & x \leq 0, \\ -\frac{1}{2}(T-t-x) & 0 \leq x \leq T-t, \\ x - (T-t) & x \geq T-t. \end{cases}$$

At point  $(t, x) = (0, 0)$ ,  $\partial_t v_1(0, 0) = \frac{1}{2}$ ,  $Dv_1(0, 0^-) = 1$ ,  $Dv_1(0, 0^+) = \frac{1}{2}$ ,  
 $D^+ v_1(0, 0) = [\frac{1}{2}, 1]$ .

## An explicit example in 1d

The dynamics:  $F_1 = [-\frac{1}{2}, 1]$ ,  $F_2 = [-1, \frac{1}{2}]$ .

The convexification:  $FC(0) = [-1, 1]$ .

The essential dynamics:  $F^E(0) = [-\frac{1}{2}, \frac{1}{2}]$ .

$$-\partial_t v_1(0,0) + \max_{p \in F^E(0)} \{-p \cdot D^+ v_1(0,0)\} = 0 \leq 0,$$

while

$$-\partial_t v_1(0,0) + \max_{p \in FC(0)} \{-p \cdot D^+ v_1(0,0)\} = \frac{1}{2} > 0.$$

The subsolution property fails for  $FC$  which is larger than  $F^E$ .

## An explicit example in 1d.

The multi-domains:  $\Omega_1 = \{x : x < 0\}$ ,  $\Omega_2 = \{x : x > 0\}$ ,  $\Gamma = \{0\}$ .

The dynamics:  $F_1 = [-\frac{1}{2}, 1]$ ,  $F_2 = [-1, \frac{1}{2}]$ .

$T > 0$  is a given final time, the final cost function  $\varphi(x) = -|x|$ .

$$v_2(t, x) := \min\{-y_{t,x}(T)\} = \begin{cases} -[x + \frac{1}{2}(T-t)] & x \geq 0, \\ -\frac{1}{2}(x + T - t) & t - T \leq x \leq 0, \\ -[x + T - t] & x \leq t - T. \end{cases}$$

At point  $(t, x) = (0, 0)$ ,  $\partial_t v_2(0, 0) = \frac{1}{2}$ ,  $Dv_1(0, 0^-) = -1$ ,  
 $Dv_1(0, 0^+) = -\frac{1}{2}$ ,  $D^-v_1(0, 0) = [-1, -\frac{1}{2}]$ .

## An explicit example in 1d

The dynamics:  $F_1 = [-\frac{1}{2}, 1]$ ,  $F_2 = [-1, \frac{1}{2}]$ .

The tangent dynamics:  $F_\Gamma(0) = \{0\}$ .

The essential dynamics:  $F^E(0) = [-\frac{1}{2}, \frac{1}{2}]$ .

$$-\partial_t v_2(0, 0) + \max_{p \in F^E(0)} \{-p \cdot D^- v_2(0, 0)\} = 0 \geq 0,$$

while

$$-\partial_t v_2(0, 0) + \max_{p \in F_\Gamma(0)} \{-p \cdot D^- v_1(0, 0)\} = -\frac{1}{2} < 0.$$

The **supersolution** property fails for  $F_\Gamma$  which is **smaller** than  $F^E$ .

## Extensions (Rao-HZ'13, Rao-Siconolfi-HZ '14)

- ▶ Weaker controllability assumptions
- ▶ Equations with Lagrangians (running cost):

$$H_i(x, p) = \sup_{\alpha \in A} \{-p \cdot f_i(x, \alpha) - \ell_i(x, \alpha)\}.$$

# References on HJB equations on multi-domains

-  Y. ACHDOU, F. CAMILLI, A. CUTRI AND N. TCHOU, *Hamilton-Jacobi equations on networks*, NODEA2013.
-  C. IMBERT, R. MONNEAU AND H. ZIDANI, *A Hamilton-Jacobi approach to junction problems and application to traffic flows*, ESAIM: COCV, 2013.
-  R.C. BARNARD AND P.R. WOLENSKI, *Flow Invariance on Stratified Domains*, Set-Valued and Variational Analysis, 2013.
-  G. BARLES, A. BRIANI AND E. CHASSEIGNE, *A Bellman approach for two-domains optimal control problems in  $\mathbb{R}^N$* , ESAIM:COCV, 2013.
-  Z. RAO, H. ZIDANI, *Hamilton-Jacobi-Bellman Equations on Multi-Domains*, Control and Optimization with PDE Constraints, Birkhäuser Basel, 2013
-  Z. RAO, A. SICONOLFI, H.ZIDANI, *Transmission conditions on interfaces for Hamilton-Jacobi-Bellman equations*, JDE 2014

# Thank you for your attention.