

Compactness estimates for Hamilton-Jacobi equations

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Partial differential equations, optimal design and numerics

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Outline

1 Hamilton-Jacobi equations

2 Compactness estimates for hyperbolic conservation laws

3 Compactness estimates for Hamilton-Jacobi equations



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Hamilton-Jacobi equations

$$\begin{cases} u_t(t, x) + H(t, x, \nabla u(t, x)) = 0 & (t, x) \in [0, T] \times \mathbb{R}^n \\ u(0, x) = u_0(x) & x \in \mathbb{R}^n \end{cases}$$

where

- $H : [0, T] \times \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ is a C^2 smooth function such that

$$(a) \quad \lim_{|p| \rightarrow \infty} \inf_{(t, x) \in [0, T] \times \mathbb{R}^n} \frac{H(t, x, p)}{|p|} = +\infty$$

$$(b) \quad D_p^2 H(t, x, p) \geq \alpha \cdot \mathbb{I}_n, \quad \forall (t, x, p) \in [0, T] \times \mathbb{R}^n \times \mathbb{R}^n$$

with $\alpha > 0$

- $u_0 : \mathbb{R}^n \rightarrow \mathbb{R}$ is a Lipschitz function

play an important role in Dynamic Optimization



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The simplest problem in the calculus of variations

Given a C^2 smooth function $L : [0, T] \times \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ such that

$$(a) \quad \lim_{|q| \rightarrow \infty} \inf_{(t,x) \in [0,T] \times \mathbb{R}^n} \frac{L(t,x,q)}{|q|} = +\infty$$

$$(b) \quad D_q^2 L(t,x,q) \geq \lambda \cdot \mathbb{I}_n, \quad \forall (t,x,q) \in [0, T] \times \mathbb{R}^n \times \mathbb{R}^n \quad (\lambda > 0)$$

consider the problem of minimizing the functional

$$J(\xi) = \int_0^T L(t, \xi(t), \xi'(t)) dt$$

over all absolutely continuous arcs $\xi : [0, T] \rightarrow \mathbb{R}^n$ satisfying

$$\xi(0) = x_0 \quad \text{and} \quad \xi(T) = x_T$$

with $x_0, x_T \in \mathbb{R}^n$



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consider the problem of **minimizing** the functional

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The action functional

A particle moving from time 0 to time T between two points $x_0, x_T \in \mathbb{R}^3$ subject to a conservative force

$$\mathbf{F}(x) = -\nabla V(x)$$

among all the (admissible) trajectories $\xi(t)$, follows the one that minimizes the action, i.e. the functional

$$J(\xi) = \int_0^T \left[\frac{1}{2} m |\xi'(t)|^2 - V(\xi(t)) \right] dt,$$

where m is the mass of the particle and $\frac{1}{2}m|\xi'(t)|^2$ is its kinetic energy



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Minimal surfaces of revolution

Given $a, b \in \mathbb{R}$ with $a < b$, consider in the space \mathbb{R}^3 the circles

$$\begin{cases} y^2 + z^2 = A^2 \\ x = a \end{cases} \quad \begin{cases} y^2 + z^2 = B^2 \\ x = b \end{cases}$$

For any smooth $\xi : [a, b] \rightarrow \mathbb{R}$, with $\xi(x) > 0$, $\xi(a) = A$ and $\xi(b) = B$, consider

- the regular curve $\vec{X}(x) = (x, 0, \xi(x))$ in the xz -plane
- the surface of revolution $\Sigma(\xi)$ generated by the rotation of \vec{X} around the x -axis

Finding the surface of revolution of minimal area amounts to minimizing

$$A(\Sigma(\xi)) = 2\pi \int_a^b \xi(x) \sqrt{1 + \xi'(x)^2} dx$$



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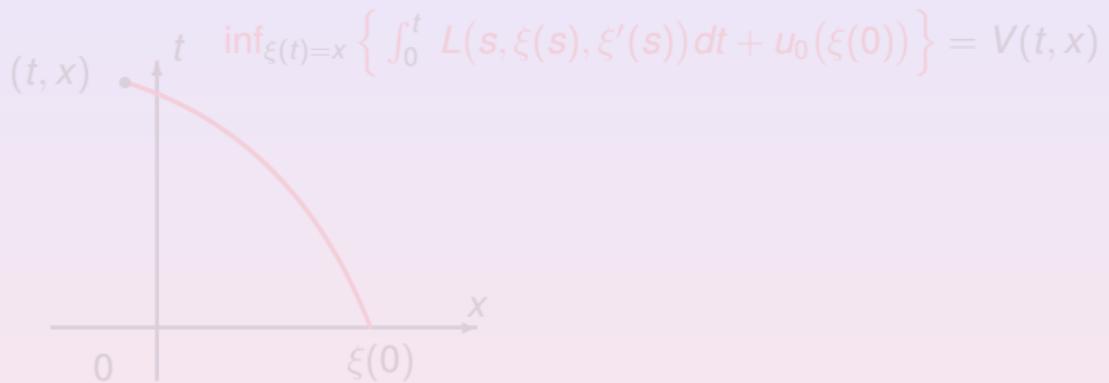
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Dynamic programming

Replacing the initial point constraint with the initial cost u_0 , consider

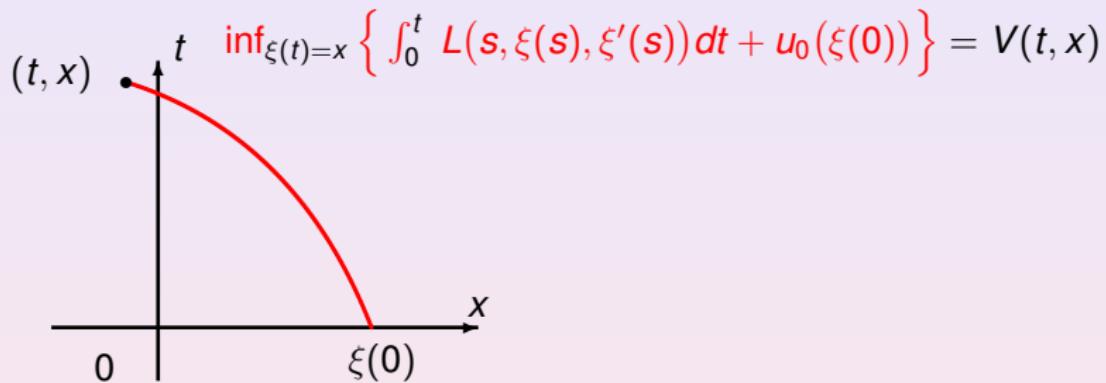


$$\begin{cases} V_t(t, x) + \underbrace{\sup_{q \in \mathbb{R}^n} \left\{ \langle q, \nabla V(t, x) \rangle - L(t, x, q) \right\}}_{H(t, x, \nabla V(t, x))} = 0 & (t, x) \in [0, T] \times \mathbb{R}^n \text{ a.e.} \\ V(0, x) = u_0(x) & x \in \mathbb{R}^n \end{cases}$$



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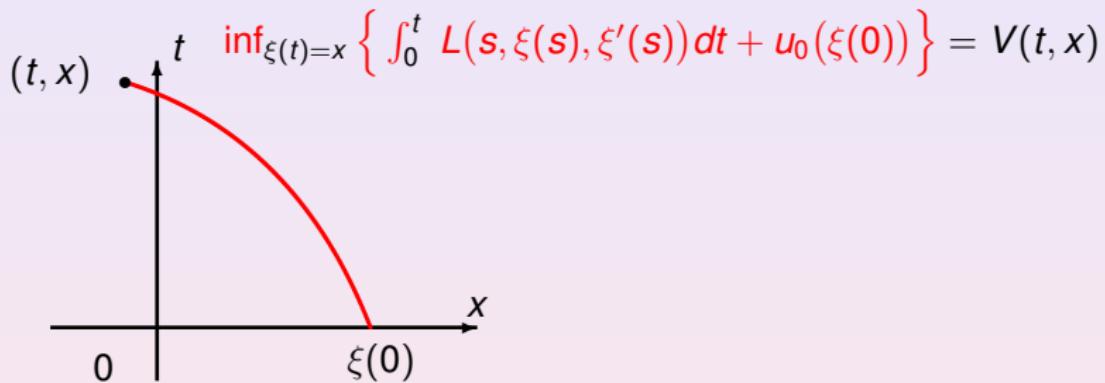


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Weak solutions to Hamilton-Jacobi equations

$$\begin{cases} u_t(t, x) + H(t, x, \nabla u(t, x)) = 0 & (t, x) \in [0, T] \times \mathbb{R}^n \\ u(0, x) = u_0(x) & x \in \mathbb{R}^n \end{cases} \quad (HJ)$$

- has no global smooth solution due to crossing of characteristics
- may have infinitely many Lipschitz solutions satisfying (HJ) a.e.
 - Dacorogna and Marcellini (1999)
- has a unique viscosity solution
 - Crandall and Lions (1983), Crandall, Evans and Lions (1984)
 - Bardi and Capuzzo Dolcetta (1997), Fleming and Soner (1993)
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Viscosity solutions

A function $u \in \mathcal{C}([0, T] \times \mathbb{R}^n)$ is a viscosity solution of

$$u_t + H(t, x, \nabla u) = 0 \quad \text{in }]; [0, T] \times \mathbb{R}^n$$

if for every $(t, x) \in (0, T) \times \mathbb{R}^n$ and every $\phi \in \mathcal{C}^1((0, T) \times \mathbb{R}^n)$

- $u - \phi$ has a local maximum at $(t, x) \Rightarrow \phi_t(t, x) + H(t, x, \nabla \phi(t, x)) \leq 0$
- $u - \phi$ has a local minimum at $(t, x) \Rightarrow \phi_t(t, x) + H(t, x, \nabla \phi(t, x)) \geq 0$



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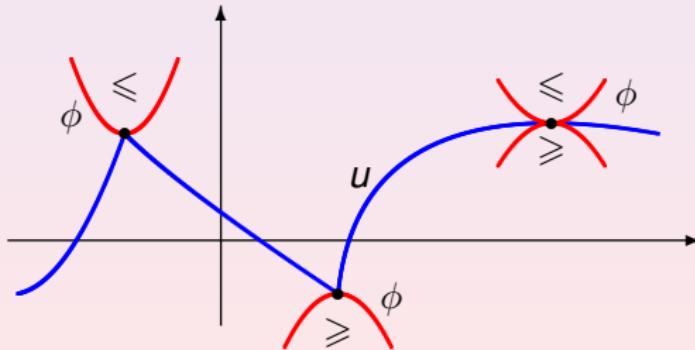
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Semicconcave functions

Definition

We say that $u : \mathbb{R}^N \rightarrow \mathbb{R}$ is (linearly) semiconcave if there exists a constant $K > 0$ (a semiconcavity constant for u) such that

$$\lambda u(x) + (1 - \lambda)u(y) - u(\lambda x + (1 - \lambda)y) \leq K\lambda(1 - \lambda)\frac{|y - x|^2}{2}$$

for all $x, y \in \mathbb{R}^N$ and all $\lambda \in [0, 1]$

- u is semiconcave with semiconcavity K if and only if the function

$$\tilde{u}(x) = u(x) - \frac{K}{2}|x|^2$$

is concave

- v is semiconvex with semiconvexity constant K if $-v$ is semiconcave with semiconcavity constant K



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For more on semiconcave functions see

- Control theory

Hrustalev (1978), C – Soner (1987), C – Frankowska (1991)

Fleming – McEneaney (2000), Rifford (2000, 2002)

- Nonsmooth and variational analysis

Rockafellar (1982)

Colombo – Marigonda (2006), Colombo – Nguyen (2010)

- Differential geometry Perelman (1995), Petrunin (2007)

- Monographs

C – Sinestrari (Birkhäuser, 2004)

Villani (Springer, 2009)



From Hamilton-Jacobi equations to conservation laws

When $n = 1$ the Hamilton-Jacobi equation

$$\begin{cases} u_t(t, x) + H(t, x, u_x(t, x)) = 0 & (t, x) \in [0, T] \times \mathbb{R} \\ u(0, x) = u_0(x) & x \in \mathbb{R} \end{cases}$$

can be reduced to the conservation law

$$\begin{cases} v_t(t, x) + H(t, x, v(t, x))_x = 0 & (t, x) \in [0, T] \times \mathbb{R} \\ v(0, x) = u'_0(x) & x \in \mathbb{R} \end{cases}$$

taking $v(t, x) = u_x(t, z)$



Scalar conservation laws

u is an **entropy solution** of

$$u_t + f(u)_x = 0 \quad \text{in } [0, +\infty) \times \mathbb{R}$$

where $f : \mathbb{R} \rightarrow \mathbb{R}$ is (uniformly) strictly convex

$$f''(u) \geq c > 0 \quad \forall u \in \mathbb{R}$$

if

- u distributional solution

$$\iint [u\varphi_t + f(u)\varphi_x] dxdt = 0 \quad \forall \varphi \in C_c^1([0, +\infty) \times \mathbb{R}) \quad (D)$$

- Lax stability condition

$$u(t, x-) \geq u(t, x+) \quad \text{for a.e } t > 0, \quad \forall x \in \mathbb{R}$$



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Compactness of the semigroup $(S_t)_{t \geq 0}$

$S_t : L^1(\mathbb{R}) \rightarrow L^1(\mathbb{R})$ associates to every initial data $u_0 \in L^1(\mathbb{R}) \cap L^\infty(\mathbb{R})$ the unique entropy solution $u(t) = S_t(u_0)$ of

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Theorem (Lax, 1954)

The map $S_t : L^1(\mathbb{R}) \longrightarrow L^1_{\text{loc}}(\mathbb{R})$ is compact for every $t > 0$

A question (by P. Lax) :

is it possible to give a quantitative estimate of the compactness of S_t ?



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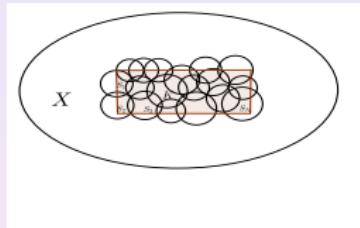
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Kolmogorov ε -entropy

Let (X, d) be a metric space and K a totally bounded subset of X



For any $\varepsilon > 0$, let $N_\varepsilon(K)$ be the minimal number of sets in a cover of K by subsets of X having diameter no larger than 2ε

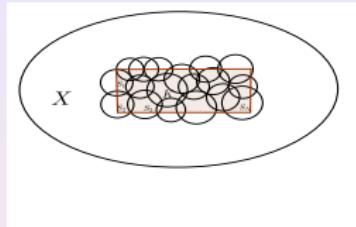
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The ε -entropy of K is defined as

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Kolmogorov ε -entropy

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Applications

one relies on Kolmogorov's ε -entropy to:

- provide estimates on the accuracy and resolution of numerical methods
- analyze computational complexity of conservation laws (derive number of needed operations to compute solutions with an error $< \varepsilon$)



Upper estimate

Given $L, m, M > 0$, define

$$\mathcal{C}_{[L,m,M]} = \left\{ u_0 \in L^1(\mathbb{R}) : \text{spt}(u_0) \subset [-L, L], \|u_0\|_{L^1} \leq m, \|u_0\|_{L^\infty} \leq M \right\}$$

Our goal: to give an upper bound for

$$\mathcal{H}_\varepsilon(S_T(\mathcal{C}_{[L,m,M]}) \mid L^1(\mathbb{R}))$$

Theorem (De Lellis and Golse, 2005)

For any $\varepsilon > 0$ and $T > 0$, one has

$$\mathcal{H}_\varepsilon(S_T(\mathcal{C}_{[L,m,M]}) \mid L^1(\mathbb{R})) \leq \frac{C_T}{\varepsilon}$$

for some constant $C_T > 0$

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For any $T > 0$ and for $\varepsilon > 0$ sufficiently small, one has

$$\mathcal{H}_\varepsilon(S_T(\mathcal{C}_{[L,m,M]}) \mid L^1(\mathbb{R})) \geq \frac{c_T}{\varepsilon}$$

for some constant $c_T > 0$

By the upper and lower bounds, we conclude

$$\mathcal{H}_\varepsilon(S_T(\mathcal{C}_{[L,m,M]}) \mid L^1(\mathbb{R})) \approx \varepsilon^{-1}$$



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Back to Hamilton-Jacobi equations

Consider the Hamilton-Jacobi equation ($n \geq 1$)

$$u_t(t, x) + H(\nabla u(t, x)) = 0 \quad (t, x) \in [0, +\infty) \times \mathbb{R}^n$$

with $H \in C^2(\mathbb{R}^n)$ satisfying

(H1) superlinearity: $\lim_{|p| \rightarrow \infty} \frac{H(p)}{|p|} = +\infty$

(H2) uniform convexity: $D^2 H(p) \geq \alpha \cdot \mathbb{I}_N, \quad \forall p \in \mathbb{R}^n$

where $\alpha > 0$ and \mathbb{I}_n is the identity $n \times n$ matrix

Legendre transform of H

$$H^*(q) = \max_{p \in \mathbb{R}^n} \{ \langle p, q \rangle - H(p) \} \quad (q \in \mathbb{R}^n)$$

is in turn superlinear and satisfies

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Hopf-Lax semigroup

For any $u_0 \in \text{Lip}(\mathbb{R}^n)$ the Cauchy problem

$$\begin{cases} u_t(t, x) + H(\nabla u(t, x)) = 0 & (t, x) \in [0, +\infty) \times \mathbb{R}^n \\ u(0, x) = u_0(x) & x \in \mathbb{R}^N \end{cases}$$

admits a unique viscosity solution given by

$$u(t, x) = \min_{y \in \mathbb{R}^n} \left\{ t \cdot H^* \left(\frac{x-y}{t} \right) + u_0(y) \right\}, \quad \forall (t, x) \in]0, +\infty[\times \mathbb{R}^n$$

Our goal: to obtain upper and lower compactness estimates for

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Semicconcavity of the Hopf-Lax semigroup

Given $K, L, M > 0$, define

$$\begin{aligned}\mathcal{C}_{[L,M]} &= \{u \in \text{Lip}(\mathbb{R}^n) : \text{spt}(u) \subset [-L, L]^n, \|\nabla u\|_{L^\infty(\mathbb{R}^n)} \leq M\} \\ \mathcal{SC}_{[K,L,M]} &= \{u \in \mathcal{C}_{[L,M]} : u \text{ semiconcave with constant } K\}\end{aligned}$$

Proposition

For any $L, M, T > 0$ and every $u \in \mathcal{C}_{[L,M]}$

- ① $S_T(u)$ is semiconcave with constant $\frac{1}{\alpha T}$
- ② $\|\nabla S_T(u)\|_{L^\infty(\mathbb{R}^n)} \leq M$
- ③ $\text{spt}(S_T(u) + T \cdot H(0)) \subset [-L_T, L_T]^n$ where $L_T = L + T \cdot \sup_{|p| \leq M} |DH(p)|$

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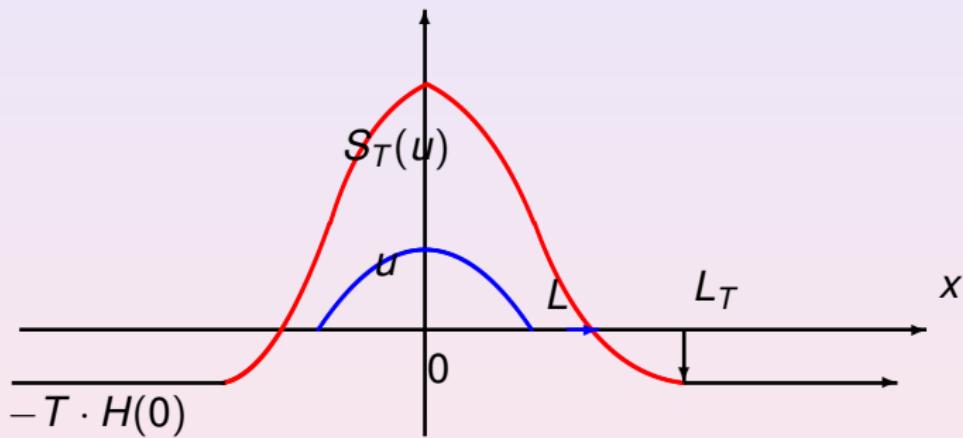
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flow associated with H-L semigroup

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For any $L, M, T > 0$ there exist constant $\varepsilon_0 = \varepsilon_0(L, M, T) > 0$ and $C = C(L, M, T) > 0$ such that, for all $\varepsilon \in (0, \varepsilon_0)$,

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Main steps of the proof

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Main ideas of the proof of the lower estimate

- 1 Controllability type result: introduce a parameterized class \mathcal{U} of smooth function and show that any element of such a class can be attained, at any given time $T > 0$, by the Hopf-Lax flow $S_T(u)$ for a suitable $u \in \mathcal{C}_{[L,M]}$
- 2 Combinatorial computation: provide an optimal (w.r.t. parameters) estimate of the maximum number of functions in \mathcal{U} that can be contained in a ball of radius 2ϵ (with respect to the norm of $W^{1,1}(\mathbb{R}^n)$)



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Reachability of semiconcave functions

Theorem

Given $K, L, M > 0$, let $T > 0$ be such that

$$K T \leq \frac{1}{2\alpha_M} \quad \text{where} \quad \alpha_M = \sup_{|p| \leq M} \|D^2 H(p)\|$$

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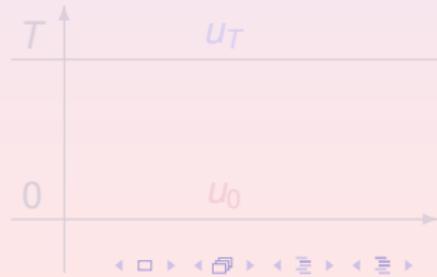
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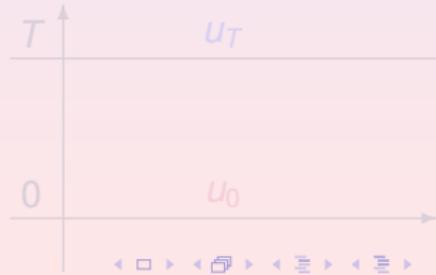
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Backward construction

Solve the equation backwards: set $v(t, x) = S_t(v_0)(x)$ with

$$v_0(x) = -u_T(-x)$$

and define

$$u(t, x) = -v(T-t, -x) \quad (t, x) \in [0, T] \times \mathbb{R}^n$$

Then

- $u(T, \cdot) = u_T$
- $u_0 \doteq u(0, \cdot) \in \mathcal{C}_{[L_T, M]}$ by the properties of S_T
- $u_t(t, x) + H(\nabla u(t, x)) = 0$ for a.e. $(t, x) \in [0, T] \times \mathbb{R}^n$

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Lower bound for $\mathcal{H}_\varepsilon \left(\mathcal{SC}_{[K,L,M]} \mid W^{1,1}(\mathbb{R}^n) \right)$

Proposition

Given $K, L, M > 0$, for any $\varepsilon > 0$

$$\mathcal{H}_\varepsilon \left(\mathcal{SC}_{[K,L,M]} \mid W^{1,1}(\mathbb{R}^n) \right) \geq \frac{\Gamma(K, L, M)}{\varepsilon^n}$$

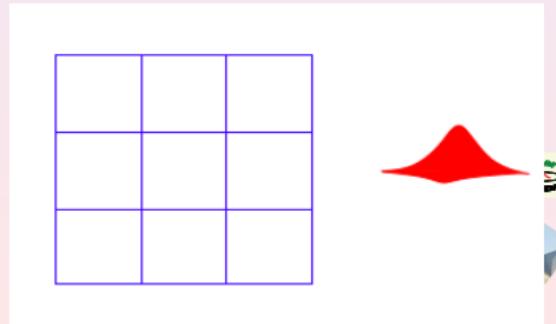
Given $N \geq 1$ integer, divide $[-L, L]^2$
into N^2 squares of side $\frac{2L}{N}$

$$[-L, L]^2 = \bigcup_{i,j=1,\dots,N} \square_{ij}$$

Construct bump functions
 $b_{ij} : \square_{ij} \rightarrow \mathbb{R}$ such that

- $\|\nabla b_{ij}\|_{\mathbb{L}^\infty} \leq \frac{KL}{12N}$, $\|b_{ij}\|_{W^{1,1}} \leq \frac{C}{N^3}$
- ∇b_{ij} Lipschitz with constant K

Sketch of the proof ($n = 2$):



The class \mathcal{U}_N of smooth functions

Let

$$\Delta_N = \left\{ \delta = (\delta_{ij})_{i,j=1}^N : \delta_{ij} \in \{-1, 1\} \right\}$$

Consider the class of smooth functions

$$\mathcal{U}_N = \left\{ u_\delta = \sum_{i,j=1}^N \delta_{ij} \cdot b_{ij} : \delta \in \Delta_N \right\}$$

Then $\#(\mathcal{U}_N) = 2^{N^2}$. Also, one can show that

- $\mathcal{U}_N \subset SC_{[K,L,M]}$
- $\|u_{\delta'} - u_\delta\|_{W^{1,1}(\mathbb{R}^2)} \leq \varepsilon$ if $\#\{(i,j) : \delta'_{ij} \neq \delta_{ij}\} \leq C_{K,L} N^{p+1} \varepsilon$

Choosing $N \approx \frac{1}{\varepsilon}$, by a combinatorial argument one can show that

$$\#\left\{\delta' \in \Delta_N : \|u_{\delta'} - u_\delta\|_{W^{1,1}(\mathbb{R}^2)} \leq \varepsilon\right\} \leq 2^{N^2} e^{-N^2/8} = e^{-N^2/8} \#(\mathcal{U}_N)$$

which yields

$$\mathcal{H}_\varepsilon\left(\mathcal{U}_N \mid W^{1,1}(\mathbb{R}^2)\right) \geq \frac{\Gamma}{\varepsilon^2}$$

with $\Gamma = \Gamma(K, L, M) > 0$. Therefore,

$$\mathcal{H}_\varepsilon\left(SC_{[K,L,M]} \mid W^{1,1}(\mathbb{R}^2)\right) \geq \frac{\Gamma}{\varepsilon^2}$$



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End of the proof of the lower estimate

want to show

Let $M > 0$ be fixed. Then, $\forall T > 0$ there exist constants $\Gamma_T > 0$ and $\Lambda_T \geq 0$ such that

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$$\Lambda_T = 2T \cdot \sup_{\|p\| \leq h} |DH(p)| \quad \text{and} \quad K_T = \frac{1}{4T|D^2H(0)|}$$

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Concluding remarks

- combining the upper and lower estimates (near $\varepsilon = 0$)

$$\mathcal{H}_\varepsilon \left(S_T(\mathcal{C}_{[L,M]}) + T \cdot H(0) \mid W^{1,1}(\mathbb{R}^n) \right) \approx \varepsilon^{-n}$$

- compactness estimates can be extended to

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(no Hopf-Lax formula available)

- reachability example of a controllability result for Hamilton-Jacobi equations



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*Thank you for your attention
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for the hospitality

