

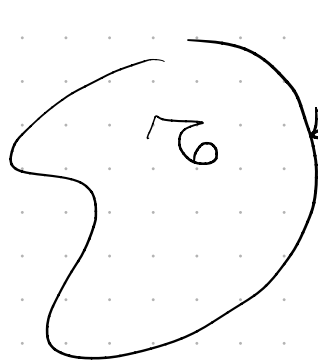
MINIMUM TIME PROBLEM

Given a dynamic

$$f: \mathbb{R}^n \times A \rightarrow \mathbb{R}^n$$

$$\textcircled{1} \begin{cases} \dot{y}(s) = f(y(s), \alpha(s)) & s \geq 0 \\ y(0) = x \end{cases}$$

Given a TARGET SET $\bar{C} \neq \emptyset \quad \bar{C} \subseteq \mathbb{R}^n$



let us define, for a given x

$$t_x(\alpha(\cdot)) = \inf \{ t \in \mathbb{R}^+ \mid y_{x,\alpha}(t) \in \bar{C} \}$$

if there exists α s.t. $y_{x,\alpha}(t) \in \bar{C}$

otherwise $t_x(\alpha) = \infty$

$$t_x(\alpha) = \begin{cases} \inf \{ t \in \mathbb{R}^+ \mid y_{x,\alpha}(t) \in \bar{C} \} \\ \infty & \text{if } y_{x,\alpha}(t) \notin \bar{C}, t > 0 \end{cases}$$

VALUE FUNCTION

$$T(x) := \inf_{\alpha \in \mathcal{A}} t_x(\alpha(\cdot))$$

the smallest over all the arrival times

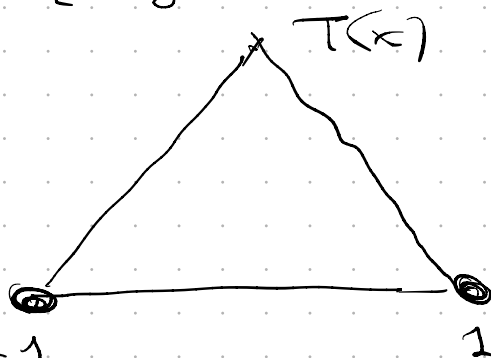
Ex $n = 1, A = [-1, 1], \bar{C} = \{-1\} \cup \{1\}$

$$f(x, a) = a$$

$$\begin{cases} \dot{y}(s) = \alpha(s) \\ y(0) = x_0 \end{cases}$$

If $x_0 \in [0, 1], \alpha^*(s) = 1$

$$1 = y^*(t) = x_0 + t \Rightarrow t = 1 - x_0$$



If $x_0 \in [-1, 0]$, $\alpha^*(s) = -1$

$$-1 = y^*(t) = x_0 - t \Rightarrow t = 1 + x_0$$

Then for $x_0 \in [-1, 1]$, $T(x) = \begin{cases} 1-x & x \in [0, 1] \\ 1+x & x \in [-1, 0] \end{cases}$

The minimum time problem is equivalent to minimize, with the same dynamic (D), the cost

$$J_x(\alpha(\cdot)) = \int_0^{t_x(\alpha)} 1 ds = t_x(\alpha)$$

If the dynamic can remain fixed at the target

$$T(x) = \inf_{\alpha \in \mathcal{A}} \int_0^{+\infty} \frac{1}{\tau_c} \mathbb{1}_{\tau_c}(y_{x,\alpha}(s)) ds =$$

$$= \inf_{\alpha \in \mathcal{A}} \left[\int_0^{t_x(\alpha)} \frac{1}{\tau_c} \mathbb{1}_{\tau_c}(y_{x,\alpha}(s)) ds + \int_{t_x(\alpha)}^{+\infty} \frac{1}{\tau_c} \mathbb{1}_{\tau_c}(y_{x,\alpha}(s)) ds \right]$$

$$\lambda = 0$$

$$L(s, \alpha(s)) = \frac{1}{\tau_c} \mathbb{1}_{\tau_c}(y_{x,\alpha}(s)) = \begin{cases} 1 & y_{x,\alpha}(s) \notin \tau \\ 0 & y_{x,\alpha}(s) \in \tau \end{cases}$$

The related HJB equation: MINIMUM TIME EQ
For a general dynamic (D) in $\mathbb{R}^n$

$$\begin{cases} \max_{\alpha \in A} [-V_\alpha(x) \cdot f(x, \alpha) - 1] = 0 & x \in \mathbb{R}^n \setminus \tau \\ V(x) = 0 & x \in \tau \end{cases}$$

Going back to the example

the minimum time equation in $(-1,1)$

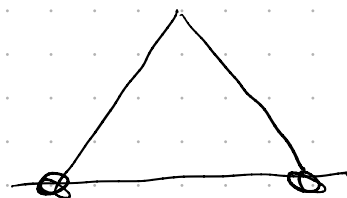
$$\begin{cases} \max_{x \in [-1,1]} [-V'(x) \alpha - 1] = 0 & x \in (-1,1) \\ V(x) = 0 & |x| = 1 \end{cases} \quad \alpha^* = \frac{-V'(x)}{|V'(x)|}$$

$$\text{Since } \max_{x \in [-1,1]} [-V'(x) \alpha - 1] = \max_{x \in [-1,1]} [-V'(x) \alpha] - 1 = |V'(x)| - 1$$

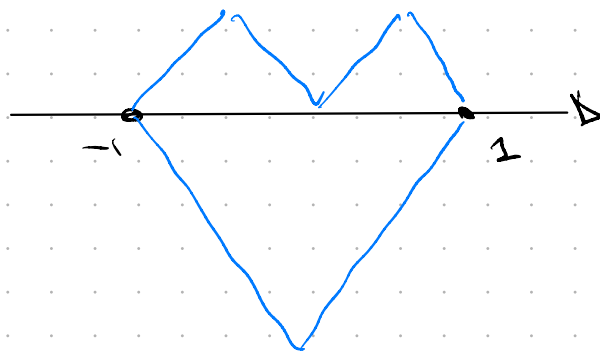
VALUE FUNCTION

the eq reads

$$\begin{cases} |V'(x)| = 1 & x \in (-1,1) \\ V(x) = 0 & |x| = 1 \end{cases} \quad (1)$$



$V(x)$ is Lipschitz and differentiable almost everywhere



One possibility to select a unique solution

is to regularize the problem adding a second

order term $-\epsilon V''$ and then pass to

the limit $\epsilon \rightarrow 0$.

$$(HJB)_\varepsilon \quad \begin{cases} -\varepsilon V''(x) + |V'(x)| = 1 & x \in (-1, 1) \\ V(-1) = V(1) = 0 \end{cases}$$

• $\forall \varepsilon > 0$ there exist a unique $V_\varepsilon \in C^2(-1, 1)$ solution to $(HJB)_\varepsilon$

Since

$$|V_\varepsilon(x) - V_\varepsilon(y)| \leq L|x-y| \quad \text{equicontinuous}$$

$$|V_\varepsilon(x)| \leq M \quad \text{uniform bound}$$

(L and M do not depend on ε)

We can pass to the limit (by Ascoli-Arzelà theorem)

$$\lim_{\varepsilon \rightarrow 0} V_\varepsilon(x) = V(x)$$

This defines the WEAK SOLUTION of the original problem (Kruskal, 60)

VISCOSITY SOLUTION

The viscosity solution V is the uniform limit of V^ε where V^ε are

$$\begin{cases} -\varepsilon \Delta V(x) + H(x, V, \nabla V) = 0 & x \in \Omega \subset \mathbb{R}^n \\ \text{b.c.} \end{cases}$$

for a given open domain $\Omega \subset \mathbb{R}^n$

$BUC(\Omega) = \{u: \Omega \subset \mathbb{R}^n \rightarrow \mathbb{R} \text{ bounded uniformly continuous}\}$

DEF Let $v \in BUC(\Omega)$ is a **VISCOSITY SOLUTION**

o f $H(x, v(x), \nabla v(x)) = 0 \quad x \in \Omega$

if and only if for any $\phi \in C^1(\Omega)$ the following conditions hold

1) at every local maximum $x_0 \in \Omega$ for $v - \phi$

$$H(x_0, v(x_0), \nabla \phi(x_0)) \leq 0$$

(i.e. v is a **VISCOSITY SUB SOLUTION**)

2) at every local minimum $x_0 \in \Omega$ for $v - \phi$

$$H(x_0, v(x_0), \nabla \phi(x_0)) \geq 0$$

(i.e. v is a **VISCOSITY SUPER SOLUTION**)

OS 1) We give a punctual definition on x_0 and we need to check for any test function

2) $\nabla \phi$ exist, but ∇v a priori does not

Going back to the example

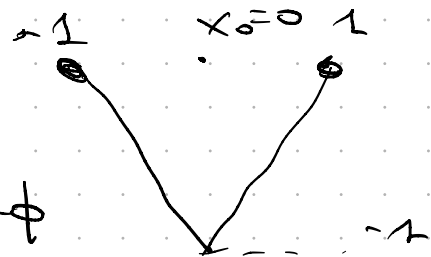
Take any solution v which has a local minimum in x_0 , choose as test function $\phi = \text{const}$

Clearly, x_0 is a minimum for $v - \phi$

$$(v - \phi)(x_0) \leq (v - \phi)(x)$$

$$|\phi'(x_0)| \geq 1 \quad \text{which is false since}$$

$$\phi = \text{const}$$



Conclusion any almost everywhere solution having a local minimum cannot be a supersolution

$$\begin{cases} |u'(x)| = 1 & x \in (-1, 1) \\ u(x) = 0 & |x| = 1 \end{cases}$$



So, $T(x)$ the minimum time value function is the UNIQUE VISCOSITY SOLUTION

PROPERTIES OF VISCOSITY SOLUTIONS

1. If $v(x)$ is a CLASSICAL solution C^1 then $v(x)$ is a VISCOSITY SOLUTION
2. If $v(x)$ is a REGULAR VISCOSITY solution then it is also CLASSICAL (i.e. it satisfies the equation pointwise)
3. the viscosity solution is the MAXIMAL SUBSOLUTION, i.e.
$$v \geq w \quad \text{for any } w \text{ sub sol}$$

CRUCIAL POINT: **UNIQUENESS**

Thm (Comparison Principle) Let
let $u, v \in BUC(\bar{\Omega})$ be respectively a sub and a supersolution s.t.

$$u(x) \leq v(x) \quad \text{for any } x \in \partial\Omega$$

then $u(x) \leq v(x)$ for any $x \in \Omega$.

VISCOSITY SOLUTION for STATIONARY FIRST

ORDER EQUATION

Given $\Omega \subseteq \mathbb{R}^n$

$$\begin{cases} H(x, v(x), \nabla v(x)) = 0 & x \in \Omega \\ v(x) = g(x) & x \in \partial\Omega \end{cases} \quad (H)$$

$H: \mathbb{R}^n \times \mathbb{R} \times \mathbb{R}^n$ is called the Hamiltonian and $g(x)$ represent a Dirichlet boundary condition

DEF $v \in BUC(\bar{\Omega})$ is a viscosity solution to (H) if it is a viscosity solution on Ω and $v(x) = g(x) \quad x \in \partial\Omega$

Ω^c
" "

EX PROBLEM with EXIT TIME and general TARGET

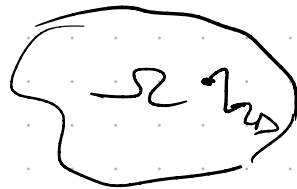
Consider Hamiltonian of the form

$$H(x, v, \nabla v) = 1 \cdot v(x) + \sup_{\alpha \in A} \{ -f(x, \alpha) \nabla v(x) \cdot l(x, \alpha) \}$$

associated with the problem of minimizing the cost functional

$$J(x) = \begin{cases} \int_0^{t_x(x)} l(y_{x,x}(s), \alpha(s)) e^{-\lambda s} ds + e^{-\lambda t_x(x)} g(y_{x,x}(t_x(x))) & \text{if } t_x(x) < \infty \\ \int_0^{\infty} l(y_{x,x}(s), \alpha(s)) e^{-\lambda s} ds & \text{if } t_x(x) = \infty \end{cases}$$

$\tau_x(\alpha)$ is the **FIRST EXIT TIME** from the open set Ω , $\bar{\Omega} = \Omega \cup \partial\Omega$



The Comparison Principle implies UNIQUENESS

Two viscosity solution u_1, u_2 .

They are both super and sub solution by the comparison principle

$$u_1 \geq u_2, \quad u_2 \geq u_1 \Rightarrow u_1 = u_2$$

The typical assumptions for EXISTENCE

- Ⓐ H uniformly continuous
 $H(x, v, \cdot)$ convex
 $H(x, \cdot, p)$ monotone

Th If Ⓐ is satisfied, $u=g \in \partial\Omega$ and the comparison principle holds then Ⓐ admits a unique viscosity solution

FINITE HORIZON CONTROL PROBLEM

Fix a terminal time T_0 and
varying the starting point of the
dynamic

$$(\tilde{D}) \quad \begin{cases} \dot{y}(s) = f(y(s), u(s)) \\ y(t) = x \end{cases} \quad \text{a general time}$$

Let $y_{x,t}(s)$ solution to $(\tilde{D})$ and
consider the cost, given T_0

$$J_{x,t}(u(\cdot)) = \int_t^{T_0} L(y_{x,t}(s), u(s)) ds + \psi(y_{x,t}(T_0))$$

for all choices of (x, t)

DEF For $x \in \mathbb{R}^n$, $t \in [0, T]$ the
VALUE FUNCTION

$$V(x, t) = \inf_{u(\cdot) \in \mathcal{U}} J_{x,t}(u(\cdot))$$

Notice that $V(x, T) = \psi(x)$ $x \in \mathbb{R}^n$

TH Assume that $V \in C^1(\mathbb{R}^n \times [0, T])$ then
 V satisfies

$$(HJB) \quad \begin{cases} -V_t(x, t) + \sup_{u \in A} \{-f(x, u) \nabla V(x, t) - L(x, u)\} = 0 \\ V(x, T) = \psi(x) \end{cases}$$

The Hamiltonian

$$H(x, \nabla \sigma) = \sup_{a \in A} \{-f(x, a) \nabla \sigma(x, t) - L(x, a)\}$$

$$\text{and (HJB)} \quad \begin{cases} -\sigma_t + H(x, \nabla \sigma(x, t)) = 0 \\ \sigma(x, T) = \psi(x) \end{cases}$$

DYNAMIC PROGRAMMING PRINCIPLE: $G > 0$

$$J(x, t) = \inf_{a \in A} \left\{ \int_t^{t+T} L(y(s), a(s)) ds + J(y(t+T), t+T) \right\}$$

FRONT PROPAGATION and MINIMUM TIME

EVOLUTIVE VS STATIONARY

The minimum time problem can be seen as an evolutive HJB by using the

~~LEVEL SET METHOD~~ great success for modeling and analysing of FRONT PROPAGATION

The front is represented by the level set of a function



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

The front propagates with normal velocity with speed $c(x)n(x)$ where $c: \mathbb{R}^n \rightarrow \mathbb{R}$ and $n: \mathbb{R}^n \rightarrow \mathbb{R}^n$ is the OUTWARD normal

Let us observe that

$$n(x) = \frac{\nabla \sigma(x, t)}{|\nabla \sigma(x, t)|}$$



$$\begin{cases} \dot{y}(s) = c(y(s)) n(y(s)) \\ y(t) = x \end{cases}$$

Then setting

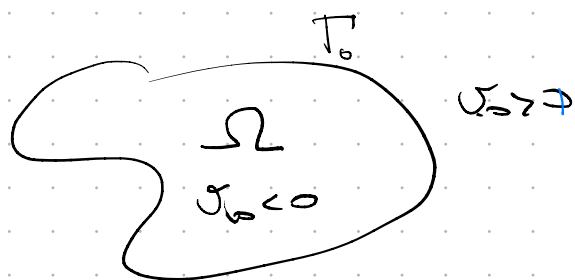
$$v(y(t), t) = 0$$

$$\begin{aligned} 0 &= \frac{d}{dt} v(y(t), t) = \nabla v(y(t), t) \cdot \dot{y}(t) + v_t(y(t), t) = \\ &= \nabla v(y(t), t) \cdot c(y(t)) \cdot m(y(t)) + v_t(y(t), t) = \\ &= \nabla v(y(t), t) \cdot c(y(t)) \cdot \frac{\nabla v(y(t), t)}{|\nabla v(y(t), t)|} + v_t(y(t), t) = \\ &= c(y(t)) |\nabla v(y(t), t)| + v_t(y(t), t) \end{aligned}$$

Since $y(t) = x$

$$\begin{cases} v_t(x, t) + c(x) |\nabla v(x, t)| = 0 \\ v(x, 0) = v_0(x) \end{cases}$$

~~FRONT~~
FRONT PROPAGATION
~~PROBLEM~~



The front propagation problem corresponds to the MINOR TIME PROBLEM with $c(x) > 0$

$$\begin{aligned} A &= B(\Omega), \quad f(x, a) = -c(x) a, \quad \Gamma_0 = \Omega \\ \ell(x, a) &= 0, \quad x = 0 \end{aligned}$$

In fact

$$\max_{a \in B(0, 1)} \{ -c(x) a \cdot \nabla T(x) \} = + c(x) |\nabla T(x)|$$

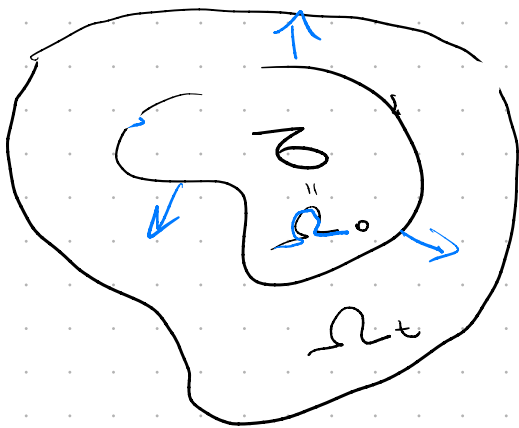
$$\uparrow a^* = - \frac{\nabla T(x)}{|\nabla T(x)|}$$

Then the minimum time problem

$$\begin{cases} |C(x)| |\nabla T(x)| = 1 & x \in \mathbb{R}^n \setminus \bar{\Omega} \\ T(x) = 0 & x \in \bar{\Omega} = \Omega \end{cases} \Leftrightarrow \begin{cases} \sigma_t(x,t) + C(x) |\nabla \sigma(x,t)| = 0 & \mathbb{R}^n \setminus (\Omega \cup \bar{\Omega}) \\ \sigma(x,0) = \sigma_0(x) \end{cases}$$

REACHABLE SET at time t : minimum time to reach $\bar{\Omega}$ from x smaller than t

$$R_t = \{x \in \mathbb{R}^n : T(x) < t\} = \{x : \sigma(x,t) < 0\} = \Omega_t$$



$$\bar{\Omega} = \{x : \sigma_0(x) = 0\}$$