

A Linear Programming Approach to Discontinuous Control Problems

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¹(joint work with Oana-Silvia Serea (CMAP))

$$\begin{cases} dX_s^{t,x,u} = b(s, X_s^{t,x,u}, u_s) ds \{ + \sigma(s, X_s^{t,x,u}, u_s) dW_s \}, & t \leq s \leq T, \\ X_t^{t,x,u} = x \in \mathbb{R}^N, \end{cases}$$

h semicontinuous,

$$V(t, x) = \inf \mathbb{E} [h(X_T^{t,x,u})],$$

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- $$\begin{cases} dx_t^{t_0, x_0, u} = b(t, x_t^{t_0, x_0, u}, u_t) dt, & t_0 \leq t \leq T, \\ x_{t_0}^{t_0, x_0, u} = x_0 \in \mathbb{R}^N, \end{cases} \quad (1)$$

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- $(0, 0) \in cl(R(T, t_0)(0, 0))$ and $(0, 0) \notin R(T, t_0)(0, 0) \implies$

$$\inf_{u \in U} h\left(x_T^{t_0, 0, 0, u(\cdot)}, y_T^{t_0, 0, 0, u(\cdot)}\right) = 1 \neq V(t_0, 0, 0).$$



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$$\begin{cases} \partial_t V(t, x) + \min_{u \in U} \langle \partial_x V(t, x), f(t, x, u) \rangle = 0, \\ \text{if } t \in (0, T), x \in \mathbb{R}^N, \end{cases} \quad (\text{HJ Mayer})$$

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Lemma

If φ is a l.s.c. supersolution of (HJ Mayer), s.t. $\varphi(T, \cdot) \geq h(\cdot)$, then

$$\varphi(t_0, x_0) \geq \inf \{ \varphi(T, x) : x \in cl(R(T, t_0) x_0) \},$$

$$\forall (t_0, x_0) \in (0, T) \times \mathbb{R}^N.$$

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Main assumptions

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- (i) b, σ bounded, uniformly continuous
- (ii) $\exists c > 0$ s.t.
 $|b(t, x, u) - b(t, y, u)| + |\sigma(t, x, u) - \sigma(t, y, u)| \leq c|x - y|$
 and
 $|b(t, x, u) - b(s, x, u)| + |\sigma(t, x, u) - \sigma(s, x, u)| \leq$
 $c|t - s|^{\frac{\delta_0}{2}},$
 $\forall (t, s, x, y, u) \in [0, T]^2 \times \mathbb{R}^{2N} \times U.$

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- $$\begin{cases} -\partial_t V_h(t, x) + H(x, DV_h(t, x), D^2 V_h(t, x)) = 0, \\ \text{for all } (t, x) \in (0, T) \times \mathbb{R}^N, & \text{(HJB)} \\ V_h(T, \cdot) = h(\cdot) \text{ on } \mathbb{R}^N, \end{cases}$$

$$H(t, x, p, A) =$$

$$\sup_{u \in \mathcal{U}} \left\{ -\frac{1}{2} \text{Tr}(\sigma \sigma^*(t, x, u) A) - \langle b(t, x, u), p \rangle \right\},$$

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- $(t, x) \in [0, T) \times \mathbb{R}^N$, $u \in \mathcal{U}$,
- $\gamma_{t,x,u}(A \times B \times C \times D) = \frac{1}{T-t} \mathbb{E} \left[\int_t^T \mathbf{1}_{A \times B \times C}((s, X_s^{t,x,u}, u_s)) ds \right] \mathbb{P}(X_T^{t,x,u} \in D)$,

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- $\int_{[t,T] \times \mathbb{R}^N \times U \times \mathbb{R}^N} (|y|^2 + |z|^2) \gamma_{t,x,u}(ds, dy, dv, dz) \leq C_0 (|x|^2 + 1)$.

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- $\gamma_{t,x,u} \in \mathcal{P}([t, T] \times \mathbb{R}^N \times U \times \mathbb{R}^N) : \forall \phi \in C_b^{1,2}([0, T] \times \mathbb{R}^N)$,

$$\int_{[t,T] \times \mathbb{R}^N \times U \times \mathbb{R}^N} \left[\begin{array}{l} (T-t) \mathcal{L}^v \phi(s, y) \\ + \phi(t, x) - \phi(T, z) \end{array} \right] \gamma(ds, dy, dv, dz) = 0$$
 (Itô's formula).

(Finite horizon) Occupational measures 2

- $J_h(t, x, u) = \mathbb{E} [h(X_T^{t,x,u})] = \int_{\mathbb{R}^N} h(z) \gamma_{t,x,u}([t, T] \times \mathbb{R}^N \times U, dz) .$

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- $\Theta(t, x) = \left\{ \begin{array}{l} \gamma \in \mathcal{P}([t, T] \times \mathbb{R}^N \times U \times \mathbb{R}^N) : \forall \phi \in C_b^{1,2}([0, T] \times \mathbb{R}^N), \\ \int_{[t,T] \times \mathbb{R}^N \times U \times \mathbb{R}^N} \begin{bmatrix} (T-t) \mathcal{L}^v \phi(s, y) \\ + \phi(t, x) - \phi(T, z) \end{bmatrix} \gamma(ds, dy, dv, dz) = 0, \\ \int_{[t,T] \times \mathbb{R}^N \times U \times \mathbb{R}^N} (|y|^2 + |z|^2) \gamma(ds, dy, dv, dz) \leq C_0 (|x|^2 + 1) \end{array} \right.$

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- $\mathcal{L}^v \phi(s, y) = \frac{1}{2} \text{Tr} [(\sigma \sigma^*)(s, y, v) D^2 \phi(s, y)] + \langle b(s, y, v), D\phi(s, y) \rangle + \partial_t \phi(s, y)$,

Linearized formulation

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- dual formulation:

$$\eta^*(t, x) = \sup_{(t, x) \in [0, T) \times \mathbb{R}^N} \left\{ \begin{array}{l} \eta \in \mathbb{R} : \exists \phi \in C_b^{1,2}([0, T] \times \mathbb{R}^N) \text{ s.t.} \\ \forall (s, y, v, z) \in [t, T] \times \mathbb{R}^N \times V \times \mathbb{R}^N, \\ \eta \leq (T - t) \mathcal{L}^v \phi(s, y) + h(z) - \phi(T, z) + \phi(t, x), \end{array} \right\},$$

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Theorem

h Lipschitz, bounded $\implies V_h(t, x) = h^*(t, x) = \eta^*(t, x),$
 $\forall (t, x) \in [0, T) \times \mathbb{R}^N.$

- $V_h(t, x) = \inf_{\gamma \in \Theta(t, x)} \int_{\mathbb{R}^N} h(z) \gamma([t, T], \mathbb{R}^N, U, dz),$
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Theorem

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- (L) inf-convolution $h_n(x) = \inf_{y \in \mathbb{R}^N} (h(y) \wedge n + n|y - x|)$
 + compactness.

What about the dual formulation? 1

- $$\begin{cases} dX_s^{t,x} = 0, \text{ for } 0 \leq t \leq s \leq T = 1, \\ X_t^{t,x} = x \in \mathbb{R}. \end{cases}, h(\cdot) = \mathbf{1}_{\{0\}}(\cdot).$$

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- $\eta^*(\frac{1}{2}, 0)$

$$= \sup \left\{ \begin{array}{l} \eta \in \mathbb{R} : \exists \phi \in C_b^{1,2}([0, 1] \times \mathbb{R}) \\ \text{s.t. } \forall (s, y, z) \in [\frac{1}{2}, 1] \times \mathbb{R}^2, \\ \eta \leq \frac{1}{2} \partial_t \phi(s, y) + h(z) - \phi(1, z) + \phi(\frac{1}{2}, 0) \end{array} \right\} = 0.$$

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- $V_h^w(t, x) = \inf_{\pi \in \mathcal{U}^w} \mathbb{E}^\pi [h(X_T^{t,x,u})]$.

Proposition

If h is u.s.c., then $V_h(t, x) = V_h^w(t, x)$, $(t, x) \in [0, T] \times \mathbb{R}^N$.

Weak control formulation. L.s.c. case

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Proposition

If convexity and h is l.s.c., then $V_h(t, x) = V_h^w(t, x)$.

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Proposition

If convexity and h is l.s.c., then $V_h(t, x) = V_h^w(t, x)$.

- **Idea of the proof:** use inf-convolution + control rules

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Thank you !