

Sensitivity analysis for HJB equations with an application to a coupled backward-forward system

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Outline

- 1 Sensitivity analysis for Hamilton-Jacobi-Bellman (HJB) equations
- 2 An application to mean field games

- \mathbf{C} : bounded continuous $f : \mathbf{R}^d \rightarrow \mathbf{R}$ with $\lim_{x \rightarrow \infty} f(x) = 0$ with sup norm.
- \mathbf{C}^1 : continuously differentiable and bounded $f : \mathbf{R}^d \rightarrow \mathbf{R}$ such that $f' \in \mathbf{C}$, $\|f\|_{\mathbf{C}^1} := \sup_x |f(x)| + \sup_x |f'(x)|$.
- \mathbf{C}^2 : twice continuously differentiable and bounded $f : \mathbf{R}^d \rightarrow \mathbf{R}$ such that $f' \in \mathbf{C}$ and $f'' \in \mathbf{C}$, $\|f\|_{\mathbf{C}^2} := \sup_x |f(x)| + \sup_x |f'(x)| + \sup_x |f''(x)|$.
- \mathbf{C}_{Lip} : Lipschitz continuous $f : \mathbf{R}^d \rightarrow \mathbf{R}$ with the norm $\|f\|_{\mathbf{C}_{Lip}} := \sup_x |f(x)| + \sup_{x,y} \frac{|f(x)-f(y)|}{|x-y|}$.
- $\mathcal{L}(\mathbf{C}^2, \mathbf{C})$: linear bounded operators from \mathbf{C}^2 to \mathbf{C} equipped with $\|\cdot\|_{\mathbf{C}^2 \rightarrow \mathbf{C}}$ and the strong operator topology.

Consider an optimal control problem

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An important example is

$\mathbf{S} = (\mathbf{C}^2)^*$, the dual space of \mathbf{C}^2

$\mathcal{M} = \mathcal{P}(\mathbf{R}^d)$, the set of probabilities on \mathbf{R}^d

$$\|\mu\|_{(\mathbf{C}^2)^*} = \sup_{\|g\|_{\mathbf{C}^2} \leq 1} \left| \int_{\mathbf{R}^d} g(x) \mu(dx) \right|.$$

$$\left\{ A[t, \mu, u] \in \mathcal{L}(\mathbf{C}^2, \mathbf{C}) : t \in [0, T], \mu \in \mathcal{M}, u \in \mathcal{U} \right\}$$

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For each $(t, \mu, u) \in [0, T] \times \mathcal{M} \times \mathcal{U}$, $A[t, \mu, u] : \mathbf{C}^2 \mapsto \mathbf{C}$ is assumed to generate a Feller process with values in \mathbf{R}^d and to be of the form

$$A[t, \mu, u]f(z) = (h(t, z, \mu, u), \nabla f(z)) + L[t, \mu]f(z)$$

- $h : [0, T] \times \mathbf{R}^d \times \mathcal{M} \times \mathcal{U} \rightarrow \mathbf{R}^d$
- $L[t, \mu] \in \mathcal{L}(\mathbf{C}^2, \mathbf{C})$ is of the form:

$$\begin{aligned} L[t, \mu]f(z) &= \frac{1}{2}(G(t, z, \mu)\nabla, \nabla)f(z) + (b(t, z, \mu), \nabla f(z)) \\ &+ \int_{\mathbf{R}^d} (f(z+y) - f(z) - (\nabla f(z), y)\mathbf{1}_{B_1}(y))\nu(t, z, \mu, dy). \end{aligned}$$

For given

$$\{u\cdot\} = \{u_t \in \mathcal{U}, t \in [0, T]\}, \quad \{\mu\cdot\} = \{\mu_t \in \mathcal{M}, t \in [0, T]\},$$

$(X_t^{\{\mu\cdot\}, \{u\cdot\}} : t \in [0, T])$ denotes the Markov process generated by $\{A[t, \mu_t, u_t]\}$ with

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Example: If $L[t, \mu] = \frac{1}{2}\sigma^2\Delta$ with a constant σ ,

$$dX_t^{\{\mu\cdot\}, \{u\cdot\}} = h(t, X_t^{\{\mu\cdot\}, \{u\cdot\}}, \mu_t, u_t) dt + \sigma dW_t.$$

The value function V is defined by

$$V(t, x; \{\mu.\}) \\ := \sup_{\{u.\}} \mathbb{E}_x \left[\int_t^T J(s, X_s^{\{\mu.\}, \{u.\}}, \mu_s, u_s) ds + V^T(X_T^{\{\mu.\}, \{u.\}}, \mu_T) \right]$$

- $J : [0, T] \times \mathbf{R}^d \times \mathcal{M} \times \mathcal{U} \rightarrow \mathbf{R}$
- $V^T : \mathbf{R}^d \times \mathcal{M} \rightarrow \mathbf{R}$

Hamilton-Jacobi-Bellman (HJB) equation

$$-\frac{\partial V}{\partial t}(t, x; \{\mu.\}) = H(t, x, \nabla V(t, x; \{\mu.\}), \mu_t) + L[t, \mu_t] V(t, x; \{\mu.\})$$
$$V(T, x; \{\mu.\}) = V^T(x; \mu_T)$$

where $H : [0, T] \times \mathbf{R}^d \times \mathbf{R}^d \times \mathcal{M} \rightarrow \mathbf{R}$ is defined by

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We aim to investigate the sensitivity of the solution $V(t, x; \{\mu.\})$ with respect to the functional parameter $\{\mu.\}$.

Assumptions:

- (A_H) on $H(t, x, p, \mu)$
- (A_{V^T}) on $V^T(x; \mu)$
- (A_L) on $\{L[t, \mu] \in \mathcal{L}(\mathbf{C}^2, \mathbf{C}) : t \in [0, T], \mu \in \mathcal{M}\}$

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- (A_U) on backward propagator $\{U_{\{\mu.\}}^{t,s} \in \mathcal{L}(\mathbf{C}, \mathbf{C}), t \leq s\}$

For any $\{\mu.\}$, $f \in \mathbf{C}^2$ and all $t \leq s \leq r$, we have

$$\frac{d}{ds} U_{\{\mu.\}}^{t,s} f = U_{\{\mu.\}}^{t,s} L[s, \mu_s] f, \quad \frac{d}{ds} U_{\{\mu.\}}^{s,r} f = -L[s, \mu_s] U_{\{\mu.\}}^{s,r} f.$$

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Smoothing property

For each $0 \leq t < s \leq T$,

$$U_{\{\mu.\}}^{t,s} : \mathbf{C} \rightarrow \mathbf{C}^1, \quad U_{\{\mu.\}}^{t,s} : \mathbf{C}_{Lip} \rightarrow \mathbf{C}^2,$$

and there exists a $\beta \in (0, 1)$ and constants $c_1, c_2 > 0$ such that

$$\|U_{\{\mu.\}}^{t,s} \phi\|_{\mathbf{C}^1} \leq c_1(s-t)^{-\beta} \|\phi\|_{\mathbf{C}}, \quad \|U_{\{\mu.\}}^{t,s} \psi\|_{\mathbf{C}^2} \leq c_2(s-t)^{-\beta} \|\psi\|_{\mathbf{C}_{Lip}}$$

for all $\phi \in \mathbf{C}$ and $\psi \in \mathbf{C}_{Lip}$.



Theorem 1 (KY2013)

Suppose $(A_H), (A_L), (A_U)$. For each fixed $\{\mu.\}$, if $V^T(\cdot; \mu_T) \in \mathbf{C}^1$, then there exists a unique mild solution V satisfying $V(t, \cdot; \{\mu.\}) \in \mathbf{C}^1$ for all $t \in [0, T]$.

For any given $\{\mu^1\}, \{\mu^2\}$, define $\bar{V} : [0, T] \times \mathbf{R}^d \times [0, 1] \rightarrow \mathbf{R}$ by

$$\bar{V}(t, x; \alpha) := V\left(t, x; \{\mu^1\} + \alpha\{(\mu^2 - \mu^1)\}\right)$$

where $\{\mu^1\} + \alpha\{(\mu^2 - \mu^1)\} := \{\mu_t^1 + \alpha(\mu_t^2 - \mu_t^1) \in \mathcal{M}, t \in [0, T]\}$.

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$$\bar{V}(t, x; 1) - \bar{V}(t, x; 0) = V\left(t, x; \{\mu^2\}\right) - V\left(t, x; \{\mu^1\}\right)$$

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$$\bar{V}(t, x; 1) - \bar{V}(t, x; 0) = V\left(t, x; \{\mu^2\}\right) - V\left(t, x; \{\mu^1\}\right)$$

Define $\bar{H}(t, x, p, \alpha) := H(t, x, p, \mu_t^1 + \alpha(\mu_t^2 - \mu_t^1))$

$$\bar{L}[t, \alpha] := L(t, \mu_t^1 + \alpha(\mu_t^2 - \mu_t^1))$$

$$\bar{V}^T(x; \alpha) := V^T(x; \mu_T^1 + \alpha(\mu_T^2 - \mu_T^1))$$

with $\alpha \in [0, 1]$, $t \in [0, T]$ and $(x, p) \in \mathbf{R}^{2d}$.

The sensitivity analysis of the solution $V(t, x; \{\mu.\})$ of

$$-\frac{\partial V}{\partial t}(t, x; \{\mu.\}) = H(t, x, \nabla V(t, x; \{\mu.\}), \mu_t) + L[t, \mu_t]V(t, x; \{\mu.\})$$

$$V(T, x; \{\mu.\}) = V^T(x; \mu_T)$$

w.r.t. a functional parameter $\{\mu.\}$ is reduced to that of the solution $\bar{V}(t, x; \alpha)$ of

$$-\frac{\partial \bar{V}}{\partial t}(t, x; \alpha) = \bar{H}(t, x, \nabla \bar{V}(t, x; \alpha), \alpha) + \bar{L}[t, \alpha]\bar{V}(t, x; \alpha)$$

$$\bar{V}(T, x; \alpha) = \bar{V}^T(x; \alpha).$$

w.r.t. a real parameter $\alpha \in [0, 1]$.

Theorem 2 (KY2013)

Assume $(A_{V^T}), (A_L), (A_U)$. Define

$$W : [0, T] \times \mathbf{R}^d \times [0, 1] \rightarrow \mathbf{R}^d, \quad W(t, x; \alpha) = \bar{U}_\alpha^{t, T} \bar{V}^T(x; \alpha).$$

Then for $\alpha_1, \alpha_2 \in [0, 1]$ with $\alpha_1 \neq \alpha_2$, there exists $c > 0$ s. t.

$$\begin{aligned} & \frac{\|W(t, \cdot; \alpha_1) - W(t, \cdot; \alpha)\|_{\mathbf{C}^1}}{|\alpha_1 - \alpha_2|} \\ & \leq c \left((T-t)^{1-\beta} \|\bar{V}^T(\cdot; \alpha_2)\|_{\mathbf{C}^2} \sup_{\substack{s \in [t, T] \\ \gamma \in [\alpha_1, \alpha_2]}} \left\| \frac{\partial \bar{L}[s, \gamma]}{\partial \alpha} \right\|_{\mathbf{C}^2 \rightarrow \mathbf{C}} \right. \\ & \quad \left. + \sup_{\gamma \in [\alpha_1, \alpha_2]} \left\| \frac{\partial \bar{V}^T(\cdot; \gamma)}{\partial \alpha} \right\|_{\mathbf{C}^1} \right) \end{aligned}$$

for every $t \in [0, T]$.

Theorem 3 (KY2013)

Assume $(A_H), (A_{V^T}), (A_L), (A_U)$. Then,

- (a) For any $T > 0$, there exists a constant $c = c(T) > 0$ such that for each $\alpha_1, \alpha_2 \in [0, 1]$ with $\alpha_1 \neq \alpha_2$,

$$\begin{aligned} & \sup_{t \in [0, T]} \frac{\|\bar{V}(t, \cdot; \alpha_1) - \bar{V}(t, \cdot; \alpha_2)\|_{\mathbf{C}^1}}{|\alpha_1 - \alpha_2|} \\ & \leq c \left(\sup_{\gamma \in [\alpha_1, \alpha_2]} \left\| \frac{\partial \bar{V}^T(\cdot; \gamma)}{\partial \alpha} \right\|_{\mathbf{C}^1} + \sup_{\substack{(t, p) \in \mathcal{O} \\ \gamma \in [\alpha_1, \alpha_2]}} \left\| \frac{\partial \bar{H}(t, \cdot, p, \gamma)}{\partial \alpha} \right\|_{\mathbf{C}} \right. \\ & \quad \left. + \sup_{\substack{t \in [0, T] \\ \gamma \in [\alpha_1, \alpha_2]}} \left\| \frac{\partial \bar{L}}{\partial \alpha}[t, \gamma] \right\|_{\mathbf{C}^2 \rightarrow \mathbf{C}} \left(\left\| \bar{V}^T(\cdot; \alpha_2) \right\|_{\mathbf{C}^2} + 1 \right) \right), \end{aligned}$$

where $\mathcal{O} = \{(t, p) : t \in [0, T], |p| \leq \sup_{t \in [0, T]} \|\bar{V}(t, \cdot; \alpha)\|_{\mathbf{C}^1}\}$.

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Assume $(A_H), (A_{V^T}), (A_L), (A_U)$. Then,

- (a) For any $T > 0$, there exists a constant $c = c(T) > 0$ such that for each $\alpha_1, \alpha_2 \in [0, 1]$ with $\alpha_1 \neq \alpha_2$,

$$\begin{aligned} & \sup_{t \in [0, T]} \frac{\|\bar{V}(t, \cdot; \alpha_1) - \bar{V}(t, \cdot; \alpha_2)\|_{\mathbf{C}^1}}{|\alpha_1 - \alpha_2|} \\ & \leq c \left(\sup_{\gamma \in [\alpha_1, \alpha_2]} \left\| \frac{\partial \bar{V}^T(\cdot; \gamma)}{\partial \alpha} \right\|_{\mathbf{C}^1} + \sup_{\substack{(t, p) \in \mathcal{O} \\ \gamma \in [\alpha_1, \alpha_2]}} \left\| \frac{\partial \bar{H}(t, \cdot, p, \gamma)}{\partial \alpha} \right\|_{\mathbf{C}} \right. \\ & \quad \left. + \sup_{\substack{t \in [0, T] \\ \gamma \in [\alpha_1, \alpha_2]}} \left\| \frac{\partial \bar{L}}{\partial \alpha}[t, \gamma] \right\|_{\mathbf{C}^2 \rightarrow \mathbf{C}} \left(\left\| \bar{V}^T(\cdot; \alpha_2) \right\|_{\mathbf{C}^2} + 1 \right) \right), \end{aligned}$$

where $\mathcal{O} = \{(t, p) : t \in [0, T], |p| \leq \sup_{t \in [0, T]} \|\bar{V}(t, \cdot; \alpha)\|_{\mathbf{C}^1}\}$.

$$\bar{V}(t, x; 1) - \bar{V}(t, x; 0) = V(t, x; \{\mu^2\}) - V(t, x; \{\mu^1\})$$

Theorem 3 (KY2013)-continued

(b) For $\{\mu^1\}, \{\mu^2\} \in C([0, T], \mathcal{M})$, there exists $k > 0$ such that

$$\sup_{t \in [0, T]} \|V(t, \cdot; \{\mu^1\}) - V(t, \cdot; \{\mu^2\})\|_{\mathbf{C}^1} \leq k \sup_{t \in [0, T]} \|\mu_t^1 - \mu_t^2\|_{\mathbf{s}}$$

and

$$\sup_{t \in [0, T]} \|\nabla V(t, \cdot; \{\mu^1\}) - \nabla V(t, \cdot; \{\mu^2\})\|_{\mathbf{C}} \leq k \sup_{t \in [0, T]} \|\mu_t^1 - \mu_t^2\|_{\mathbf{s}}$$

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Backward HJB equation

$$-\frac{\partial V}{\partial t}(t, x; \{\mu.\}) = H(t, x, \nabla V(t, x; \{\mu.\}), \mu_t) + L[t, \mu_t] V(t, x; \{\mu.\})$$

$$V(T, x; \{\mu.\}) = V^T(x; \mu_T)$$

$$H(t, x, p, \mu) = \max_{u \in \mathcal{U}} (h(t, x, \mu, u)p + J(t, x, \mu, u))$$

Let $\hat{u}(t, x, \{\mu.\})$ denote the unique optimal control function.

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Regularity condition of feedback control

For every $\{\eta.\}, \{\xi.\} \in C_{\mu_0}([0, T], \mathcal{P}(\mathbf{R}^d))$,

$$\sup_{(t,x) \in [0,T] \times \mathbf{R}^d} |\hat{u}(t, x; \{\eta.\}) - \hat{u}(t, x; \{\xi.\})| \leq k_1 \sup_{s \in [0, T]} ||\eta_s - \xi_s||_{(\mathbf{C}^2)^*}$$

with some constant $k_1 > 0$.

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$$V(T, x; \{\mu.\}) = V^T(x; \mu_T)$$

$$H(t, x, p, \mu) = \max_{u \in \mathcal{U}} (h(t, x, \mu, u)p + J(t, x, \mu, u))$$

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Forward kinetic equation

$$\frac{d}{dt}(g, \mu_t) = (A[t, \mu_t, \hat{u}(t, ., \{\mu.\})]g, \mu_t), \quad \mu_0 = \mu$$

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