

Programming dynamic game:  $0 \leq t_1 \leq t_2 \leq T$

$$v(t_1, x) = \text{Inf}_{\alpha \in \mathcal{A}(t_1)} \mathbb{E} \left[ \int_{t_1}^{t_2} f(s, X_s^{t_1, x, \alpha}, \alpha_s) ds + v(t_2, X_{t_2}^{t_1, x, \alpha}) \right]$$

[ Lemma:  $v$  is  $\frac{1}{2}$  Hölder in  $t$  and  $\alpha$  in  $x$  ]

Proof: postponed.

Theo (DPP 2). If  $\tau$  stopping time between  $t$  and  $T$ , then

$$v(t, x) = \text{Inf}_{\alpha \in \mathcal{A}(t)} \mathbb{E} \left[ \int_t^{\tau} f(s, X_s^{t, x, \alpha}, \alpha_s) ds + v(\tau, X_{\tau}^{t, x, \alpha}) \right]$$

[Not proved]

Heuristic for the PDE: if  $v \in C_b^{1,2}$

for  $\tau = t + h$  ( $h > 0$  small)

$$\begin{aligned} \circ \cancel{v}(t, x) &= \text{Inf}_{\alpha \in \mathcal{A}(t)} \mathbb{E} \left[ \int_t^{t+h} f ds + \cancel{v}(t, x) \right. \\ &+ \int_t^{t+h} (\partial_t v + b(s, X_s, \alpha_s) \cdot \nabla v(s, X_s) \\ &+ \frac{1}{2} \text{Tr}(\sigma \sigma^T(s, X_s, \alpha_s) D^2 v(s, X_s))) ds \\ &\left. + \int_t^{t+h} \nabla v(s, X_s) \cdot \cancel{\sigma}(s, X_s, \alpha_s) dW_s \right] \end{aligned}$$

martingale

As  $h$  small,  $s \approx t$  &  $X_s \approx x$  on  $(t, t+h]$

$$\text{So } 0 \approx \inf_{x \in \mathcal{J}(t)} \mathbb{E} \left[ \int_t^{t+h} (f_t \circlearrowleft u + b \cdot \nabla u + \frac{1}{2} \text{Tr}(\sigma \sigma^T D^2 u)) (t, x, \alpha_s) ds \right] + o(h)$$

$$0 \stackrel{''}{=} \inf_{a \in A} \mathbb{E} \left[ \int_t^{t+h} (f_t \circlearrowleft u + b \cdot \nabla u + \frac{1}{2} \text{Tr}(\sigma \sigma^T D^2 u)) (t, x, a) ds \right] + o(h)$$

$$\Rightarrow 0 = \inf_{a \in A} \left\{ f_t \circlearrowleft u + b \cdot \nabla u + \frac{1}{2} \text{Tr}(\sigma \sigma^T D^2 u) \right\} (t, x, a)$$

let  $h \rightarrow 0$  ↪

$$(HJ) \quad \left\{ \begin{array}{l} - \partial_t u(t, x) + \sup_{a \in A} \left\{ \begin{array}{l} - b(t, x, a) \cdot \nabla u(t, x) \\ - \frac{1}{2} \text{Tr}(\sigma \sigma^T(t, x, a) D^2 u(t, x)) \\ - f(t, x, a) \end{array} \right\} = 0 \\ u(T, x) = g(x) \text{ in } \mathbb{R}^d \end{array} \right. \text{ in } (0, T) \times \mathbb{R}^d$$

Hamilton-Jacobi eq.

Hamiltonian:

$$H(t, x, p, A) = \sup_{a \in A} \left\{ \begin{array}{l} - b \cdot p - \frac{1}{2} \text{Tr}(\sigma \sigma^T A) \\ - f \end{array} \right\}$$

and HJ becomes

$$(HJ) \quad \left\{ \begin{array}{l} - \partial_t u + H(t, x, \nabla u, D^2 u) = 0 \text{ in } (0, T) \times \mathbb{R}^d \\ u(T, x) = g(x) \text{ in } \mathbb{R}^d \end{array} \right.$$

## Interest of HJ eq

- ① Numerical reason: "finite difference method" to compute the sol of HJ.
- ② Analytic tools to analyse  $J$ ,  $!$  and regularity of the sol of HJ
- ③ Verification principle (optimal synthesis).

### Theo (Verification principle)

① Assume that  $v: [0, T] \times \mathbb{R}^d \rightarrow \mathbb{R}$  of class  $C^{1,2}$  in  $(0, T) \times \mathbb{R}^d$ , continuous,  $b_d$ , in  $[0, T] \times \mathbb{R}^d$  sol of HJ, then

$$v \leq u$$

② Assume in addition that there is  $\alpha^*: [0, T] \times \mathbb{R}^d \rightarrow A$ , Lipsch. in  $x$ , indep of  $t$ , such that

$$\begin{aligned} H(t, x, Dv(t, x), D^2v(t, x)) = \\ - b(t, x, \alpha^*(t, x)) \cdot Dv(t, x) - \frac{1}{2} \text{Tr} \left( \begin{matrix} \sigma^T(t, x, \alpha^*(t, x)) \\ \times D^2v(t, x) \end{matrix} \right) \\ - f(t, x, \alpha^*(t, x)). \end{aligned}$$

Then  $v = v^*$  and  $\alpha^*$  is optimal. For any  $t, u$ ,  
 Meaning of " $\alpha^*$  optimal": let

$$X_s^* \text{ the sol to } \begin{cases} dX_s = b(s, X_s, \alpha^*(s, X_s)) ds \\ \quad + \sigma(s, X_s, \alpha^*(s, X_s)) dW_s \\ X_t = x \end{cases}$$

Then  $v(t, u) = J(t, u, \alpha^*(\cdot, X^*(\cdot)))$ .

$\alpha^* = \alpha^*(t, u)$  is called feedback.

Ex:  $\begin{cases} dX_t = a dt + dW_t^A & d=1 \\ X_t = u. \end{cases} \quad A = \mathbb{R}$

$$v(t, u) = \inf_{\alpha \in \mathcal{A}(H)} \mathbb{E} \left[ \int_t^T \frac{1}{2} |\alpha_s|^2 ds + g(X_T) \right]$$

where  $g$  Lipschitz  $\leq$  bd.

$$H(t, x, p, A) = \sup_{a \in \mathbb{R}} \left\{ -a \cdot p - \frac{1}{2} A - \frac{1}{2} a^2 \right\}. \quad \begin{aligned} b(s, x, a) &= a \\ \sigma &= \frac{1}{2} a^2 \end{aligned}$$

$$= -\frac{1}{2} A + \sup_{a \in \mathbb{R}} \left\{ -a \cdot p - \frac{1}{2} a^2 \right\}$$

$$= -\frac{1}{2} A + \frac{p^2}{2} \quad \text{for } \bar{\alpha}(t, x, p) = -p.$$

HJ eq

$$(HJ) \begin{cases} -\partial_t v - \frac{1}{2} \partial_{xx} v + \frac{1}{2} |\partial_x v|^2 = 0 \\ v(T, x) = g(x) \end{cases}$$

If  $v$  is a smooth sol of (HJ), then  
 $\alpha^*(t, x) = -\partial_x v(t, x)$ .

Proof of the verification theorem: (Hyp  $v \in C_b^{1,2}$ )

① Fix  $(t, x)$ , let  $\alpha \in \mathcal{A}(H)$ .

$$\begin{aligned} g(X_T) &= v(T, X_T^{t,x,\tau}) \quad (X_s = X_s^{t,x,\tau}) \\ &= v(t, x) + \int_t^T \left( \partial_t v + b \cdot \partial_x v + \frac{1}{2} \text{Tr}(\sigma \sigma^T D_x^2 v) \right) ds \\ &\quad + \int_t^T \partial_x v \cdot \sigma dW_s \end{aligned}$$

where all terms are computed at  
 $(s, X_s, \tau_s)$

$$\text{and where } -\partial_t v - b \cdot \partial_x v - \frac{1}{2} \text{Tr}(\sigma \sigma^T D_x^2 v) - f \leq -\partial_t v + H(t, x, \partial_x v, D_x^2 v) = 0.$$

$$\Rightarrow \mathbb{E} \left[ \underline{g}(X_T) \right] \geq v(t, x) + \mathbb{E} \left[ \int_t^T -f(s, X_s, \tau_s) ds \right].$$

$$\Rightarrow v(t, x) \leq \inf_{\alpha \in \mathcal{A}(H)} J(t, x, \alpha) = v(t, x).$$

② Let  $X^*$  solve the SDE

$$\begin{cases} dX_s = b(s, X_s, \alpha^*(s, X_s)) ds \\ \quad + \sigma(s, X_s, \alpha^*(s, X_s)) dW_s \\ X_T = u \end{cases} \quad (\exists! \text{ sol})$$

let  $\bar{\alpha}_s = \alpha^*(s, X_s)$

Then  $X_s^* = X_s^{t, x, \bar{\alpha}} \quad \forall s \in [t, T]$ .

Using  $\bar{\alpha}$  in the computation above

Then  $v(t, x) = J(t, x, \bar{\alpha}) \stackrel{(\geq v(t, x))}{\leq} v(t, x)$   
 $\uparrow$   
 previous proof

$\Rightarrow \bar{\alpha}$  optimal for  $v(t, x)$

and  $v(t, x) = J(t, x, \bar{\alpha}) = v(t, x)$ .

Exo Compute  $v(t, x)$  for the 1d example when  $f$  quadratic and justify the verification thesis in this setting.



