

Stochastic Diff. Game

(1) Two players games:

Def: If we have 2 players on action

sets A, B and objective function

$$F: A \times B \rightarrow \mathbb{R}.$$

i) It's zero-sum game if A is

to maximize F and B is to max

-imize $-F$

ii) Nash equi. in such zero-sum

game is $(a^*, b^*) \in A \times B$ s.t.

$$\inf_b F(a^*, b) = F(a^*, b^*) = \sup_a F(a, b^*).$$

Proof: If such Nash equilibrium

exists. Then, we have:

It's called $\leftarrow F(a^*, b^*) = \sup_a \inf_b F(a, b) = \inf_b \sup_a F(a, b)$
value of
the game

Since $\inf_b F(a^*, b) \leq \sup_a D \dots \sup_a \dots$

i.e. it means Nash equi. is a
saddle point of F .

Next, we suppose the two players α, β .

control a k -dim process X :

$$dX_t = b(X_t, \tau_t, \beta_t) dt + \sigma(X_t, \alpha_t, \beta_t) dW_t.$$

And the objective is:

$$J(\alpha, \beta) := \mathbb{E} \left(\int_0^T f(X_t, \alpha_t, \beta_t) dt + g(X_T) \right).$$

Where $\alpha, \beta: (0, T) \times \mathbb{R}^k \rightarrow A, B$, are Markovian

control. / closed loop control. (i.e. $\tau = \tau(t, X_t)$.)

Remark: i) If in the case of one player

there's no difference whether we

use Markovian control or open

control. (measurable). Which can be

represented by $(t, X_t(w))$.

(ii) There's another kind Markovian

control: $\tau(w) = \alpha(t, (X_t(w))_t)$.

which's path-dependent

Def: The game has value if $\sup_{\alpha} \inf_{\beta} J(\alpha, \beta)$

$$= \inf_{\beta} \sup_{\alpha} \bar{J}(\alpha, \beta).$$

To optimize: Fix β, α . resp. we can get

V^{β}, V^{α} satisfy HJB:

$$\partial_t V^{\beta} + \sup_{\alpha} h(x, \nabla V^{\beta}, \nabla^2 V^{\beta}, \alpha, \beta) = 0$$

$$\partial_t V^{\alpha} + \inf_{\beta} h(x, \nabla V^{\alpha}, \nabla^2 V^{\alpha}, \alpha, \beta) = 0$$

where $h(x, \eta, z, \alpha, \beta) = b(x, \alpha, \beta) \cdot \eta$

$$+ \frac{1}{2} \text{Tr}(\sigma \sigma^T(x, \alpha, \beta) \cdot z) + f(x, \alpha, \beta).$$

$$\text{and } V^{\alpha}(T, x) = V^{\beta}(T, x) = g(x).$$

Remark: If (α^*, β^*) is Nash. Then: V^{α^*}

V^{β^*} both satisfy the same PDE:

$$\partial_t V + h(x, \nabla V, \nabla^2 V, \alpha^*, \beta^*) = 0.$$

\Rightarrow By Feynman-Kac representation:

$$\text{we have } V = V^{\alpha^*} = V^{\beta^*}.$$

$$\text{And } \sup_{\alpha} h(\dots) = \inf_{\beta} h(\dots), \text{ i.e.}$$

(α^*, β^*) is saddle point of h .

Def. i) $M^+(x, \eta, z) := \inf_{\beta} \sup_{\alpha} h(x, \eta, z, \alpha, \beta).$

$$M^-(x, \eta, z) := \sup_{\alpha} \inf_{\beta} h(x, \eta, z, \alpha, \beta).$$

ii) We say Isaacs's condition holds

$$\text{if } M^+ \equiv M^-.$$

Remark: i) By Remark above, Nash equilibrium

exists \Rightarrow Isaacs's condition holds.

ii) Isaacs's condition is key to

ensure value exists.

Thm. (Verification)

If Isaacs's condition holds ($M \triangleq M^\pm$).

and $\exists V \in C^1$. st.

$$\partial_t V(t, x) + H(x, \nabla V, \nabla^2 V) = 0. \quad V(T, x) = g(x).$$

(α^*, β^*) is measurable saddle point of

$$h(x, \nabla V, \nabla^2 V, \alpha, \beta). \quad \forall (t, x).$$

Prisus, the state equation $\dot{X}_t =$

$$b(x_t, \alpha(t, x_t), \beta(\dots)) \dot{A}_t + \dots$$
 is well-posed.

Then: (α^*, β^*) is Nash equi. closed loop.

Ex. (Buck Rahmi's)

When we use open loop control.

the value may not exist.

Consider $X = (X^1, X^2)$. $W = (W^1, W^2)$.

$$\begin{cases} \dot{X}_t^1 = \alpha A_t + \sigma A W_t^1 & X_0^1 = X_0^2 \\ \dot{X}_t^2 = \beta A_t + \sigma A W_t^2 & = X_0. \end{cases}$$

the value span of control is $A=B=[0,1]$.

Finite objective $J(\alpha, \beta) = \mathbb{E}(|X_T^1 - X_T^2|)$.

$$\Rightarrow h(\eta^1, \eta^2, z, \alpha, \beta) = \alpha \eta^1 + \beta \eta^2 + \frac{1}{2} \sigma^2 (z_1 + z_2).$$

$$\text{And } H^{\pm}(x, \eta, z) = |\eta^1| - |\eta^2|.$$

So it satisfies Isaacs's condition. But

$$\underline{\text{Thm:}} \quad \underline{V}_0 := \sup_{\alpha} \inf_{\beta} J < \overline{V}_0 := \inf_{\beta} \sup_{\alpha} J.$$

if $0 \leq \sigma < \frac{1}{2} \sqrt{T}$, in open loops.

$$\underline{\text{Pf:}} \quad 1) \text{ Set } \alpha = \beta \Rightarrow J(\alpha, \beta) = 2\sigma \sqrt{T/2}$$

$$\underline{\text{S:}} \quad \underline{V}_0 \leq 2\sigma \sqrt{T/2}.$$

$$2) \text{ Set } \tilde{\alpha} = - \frac{\mathbb{E}(X_T^2)}{|\mathbb{E}(X_T^1)|} \quad \begin{array}{l} \mathbb{I} \in \mathbb{E}(X_T^1 \neq 0) + \\ \mathbb{I} \in \mathbb{E}(X_T^1 = 0) \end{array}$$

$$\Rightarrow J(\tilde{\alpha}, \beta) \geq |\mathbb{E}(X_T^1 - X_T^2)|.$$

$$= |\tilde{\alpha} T + \tilde{\alpha} \mathbb{E}|X_T^1|| \geq T.$$

$$\underline{\text{S:}} \quad \overline{V}_0 \geq T.$$

c2) n-player game:

① Consider n players choose control

$(\tau_i^j)^n \in A_1 \times \dots \times A_n$ to influence:

$$\dot{X}_t = b(X_t, \vec{\tau}_t) dt + \sigma(X_t, \vec{\tau}_t) dW_t$$

$$b, \sigma: \mathbb{R}^k \times \prod_{i=1}^n A_i \rightarrow \mathbb{R}^k, \mathbb{R}^{k \times m}$$

Next, we consider the closed loop

control to optimize objective:

$$J_i(\vec{\alpha}) := \mathbb{E} \left[\int_0^T f_i(X_t, \vec{\tau}(t, X_t)) dt + g_i(X_T) \right]$$

Def: i) closed loop Nash eqn:

$\vec{\alpha}$ is: $\forall i, \forall \beta \in A_i$, we have:

$$J_i(\vec{\alpha}) \geq J_i(\vec{\alpha}^{-i}, \beta)$$

ii) Hamilton for player i is $h_i(x, \eta, z, \vec{\alpha}) = b(x, \vec{\alpha}) \cdot \eta + \frac{1}{2} \text{Tr}(\sigma \sigma^T(x, \vec{\alpha}) z) + f_i(x, \vec{\alpha})$

RMK: Each player will have
 their own value func.
 $v_i(t, x)$, as well!

iii) generalized Isaacs' condition holds
 if \exists measurable $q_i: [0, T] \times \mathbb{R}^d \times C(\mathbb{R}^d)^n$
 $\times S_i^n \rightarrow A_i$, st. $\vec{q} := (q^1, \dots, q^n)$ is
 Nash equi for static n -player
 game with given func. $\prod_i A_i \ni (a_1, \dots, a_n)$
 $\mapsto h_i(x, q_i, z_i, a_1, \dots, a_n)$, $\forall (x, \vec{q}, \vec{z})$.

Theorem 7.3 (Verification theorem). Suppose the generalized Isaacs' condition holds. Suppose $\vec{v} = (v_1, \dots, v_n)$, with $v_i: [0, T] \times \mathbb{R}^d \rightarrow \mathbb{R}$ for each i , is a $C^{1,2}$ solution of the PDE system

$$\partial_t v_i(t, x) + h_i\left(x, \nabla v_i(t, x), \nabla^2 v_i(t, x), \vec{\alpha}(x, \nabla \vec{v}(t, x), \nabla^2 \vec{v}(t, x))\right) = 0,$$

$$v_i(T, x) = g_i(x),$$

where we abbreviate $\nabla \vec{v} = (\nabla v_1, \dots, \nabla v_n)$ and $\nabla^2 \vec{v} = (\nabla^2 v_1, \dots, \nabla^2 v_n)$. Finally, setting $\vec{\alpha}^*(t, x) = \vec{\alpha}(x, \nabla \vec{v}(t, x), \nabla^2 \vec{v}(t, x))$, suppose that the state equation

$$dX_t = b(X_t, \vec{\alpha}^*(t, X_t))dt + \sigma(X_t, \vec{\alpha}^*(t, X_t))dW_t$$

is well-posed. Then $\vec{\alpha}^*$ is a closed loop Nash equilibrium.

Pf: Note $\partial_t V_i + \sup_{A_i} h_i(t, x, \nabla V_i, \dots) = 0$

by def of $\vec{\pi}$. (Issacs's cond.)

Apply Verification Thm on one

player. $\Rightarrow V_i(t, x) = \sup_{\alpha_i} \mathbb{E} \left(\int_t^T \right.$

$f_i(t, x, \vec{\pi}^i(s, X_s), \alpha_i) dt + g_i(X_T) \Big)$

where $\mu X_s^{t,x} = b_i(t, x, \vec{\pi}^i(s, X_s), \alpha_i)$

$\alpha_i) dt + \sigma_i(\dots) dW_s$.

By Issacs's cond. $\Rightarrow \vec{\pi}^i$ is point-

wise optimizer. So it's truly op-

timal control. Repeat on i .

② Private State process:

Consider $n = m + 1$. And dynamics:

$\mu X_t^i = b_i(\vec{X}_t, \alpha_t^i) dt + \sigma_i(\vec{X}_t, \alpha_t^i) dW_t + \hat{\sigma}_i$

Set objective is: $(\vec{X}_t) dB_t$

$J_i(\vec{\pi}) = \mathbb{E} \left(\int_0^T f_i(\vec{X}_t, \alpha_t^i) dt + g_i(\vec{X}_T) \right)$

Set the Hamilton of i^{th} is

$$h_i(x, y, z, \vec{a}) = \sum_1^{\hat{n}} b_k(x, a_k) y_k + \frac{1}{2} \sum_1^{\hat{n}} \sigma_k^2(x, a_k) z_k^2 + \frac{1}{2} \sum_{k \neq j}^{\hat{n}} \tilde{\sigma}_k(x) \tilde{\sigma}_j(x) z_{kj} + f_i(x, a_i)$$

Knok: Set $\tilde{h}_i(x, y, z, a_i) = b_i(x, a_i) y_i + \frac{1}{2} \sigma_i^2(x, a_i)$

$$\Rightarrow h_i - \tilde{h}_i(\dots, a_i) \stackrel{\Delta}{=} h_i' \quad z_{ii} + f_i(x, a_i)$$

is indapt of a_i .

Lemma. For each i , if \exists measurable κ_i

$$: [0, T) \times \mathbb{R}^n \times \mathbb{R} \times \mathbb{R} \rightarrow A_i, \text{ st. } \forall (x, y, z).$$

We have: $\kappa_i(x, y, z) \in \operatorname{argmax}_{A_i} \tilde{h}_i(\dots, a_i)$

\Rightarrow Isaacs's condition holds.

\Rightarrow Set optimize Hamilton $H_i(x, y, z) = \sup_{A_i}$

We have HJB system: $\tilde{h}_i(\dots, a_i)$

$$0 = \partial_t V_i(t, x) + H_i(x, \partial_{x_i} V_i, \partial_{x_i} V_i) + h_i'$$

$$V_i(t, x) = g_i(x), \text{ for } \alpha_i^* \quad (x, \partial_{x_i} V_i \dots)$$