

# Set Values and Efficiency of Nonzero Sum Games

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# Outline

- 1 Introduction
- 2 Efficiency and mechanism design
- 3 Dynamic set value of games

# Nash equilibrium

- Consider a nonzero sum game with two players :
  - ◇ controls :  $a = (a_1, a_2) \in A = A_1 \times A_2$
  - ◇ utilities :  $J(a) = (J_1(a), J_2(a))$
- **Nash Equilibrium (NE)** :  $a^* \in A$

$$J_1(a^*) \geq J_1(a_1, a_2^*), \quad J_2(a^*) \geq J_2(a_1^*, a_2), \quad \forall a_1, a_2$$

# Ill-posedness

- An example :

$J(a)$	$a_2 = 0$	$a_2 = 1$	$a_2 = 2$
$a_1 = 0$	(100, 100)	(0, 102)	(0, 102)
$a_1 = 1$	(102, 0)	(1, 2)	(0, 0)
$a_1 = 2$	(102, 0)	(0, 0)	(3, 1)

- **Multiple NE** : (raw) set value  $\mathbb{V}_0 := \{(1, 2), (3, 1)\}$
- **Inefficiency** : NEs  $<$  socially optimal (100, 100)
- **Instability** : A small perturbation of the game may change the efficiency dramatically

# Two goals

- **Mechanism design** : Improve the efficiency by **small** "investment"
- **Dynamic set value** :
  - ◇ DPP
  - ◇ PDE approach

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# Efficiency

- Recall the example

$J(a)$	$a_2 = 0$	$a_2 = 1$	$a_2 = 2$
$a_1 = 0$	(100, 100)	(0, 102)	(0, 102)
$a_1 = 1$	(102, 0)	(1, 2)	(0, 0)
$a_1 = 2$	(102, 0)	(0, 0)	(3, 1)

- Efficiency =  $\frac{\text{best equilibrium}}{\text{socially optimal control}} = \frac{3 + 1}{100 + 100} = 2\%$ 
  - Price of stability (Anshelevich, Dasgupta, et al 2008)
  - Relatively easy to implement the best equilibrium

## Mechanism 1 : $\kappa$ -implementation

- $\kappa$ -implementation (Monderer-Tennenholtz 2003) :
  - ◇ A mediator designs a rewarding mechanism  $\pi = (\pi_1, \pi_2) \in \Pi_\kappa$

$$\pi_i \geq 0, \quad \pi_1 + \pi_2 \leq \kappa$$

- ◇ Consider the modified game :  $J^\pi(a) := J(a) + \pi(a)$
- The focus of Monderer-Tennenholtz 2003 :
  - ◇ Find minimum  $\kappa$  to induce a desired outcome
  - ◇ ...
- Our focus : improvement of efficiency by small  $\kappa$

$$E_\kappa := \sup_{\pi \in \Pi_\kappa} E(\pi)$$



## $\kappa$ -implementation : an example

- $\kappa = 1$ , set  $\pi(0, 1) = (1, 0)$  and  $\pi(a) = (0, 0)$  for all other  $a$

$J^\pi(a)$	$a_2 = 0$	$a_2 = 1$	$a_2 = 2$
$a_1 = 0$	(100, 100)	(1, 102)	(0, 102)
$a_1 = 1$	(102, 0)	(1, 2)	(0, 0)
$a_1 = 2$	(102, 0)	(0, 0)	(3, 1)

◇  $E(\pi) = \frac{0+102}{100+100} = 51\%$

- $\kappa = 4$ , set  $\pi(0, 0) = (2, 2)$  and  $\pi(a) = (0, 0)$  for all other  $a$

$J^\pi(a)$	$a_2 = 0$	$a_2 = 1$	$a_2 = 2$
$a_1 = 0$	(102, 102)	(0, 102)	(0, 102)
$a_1 = 1$	(102, 0)	(1, 2)	(0, 0)
$a_1 = 2$	(102, 0)	(0, 0)	(3, 1)

◇  $E(\pi) = 100\%$

# The efficiency function

- The efficiency function

$$E_{\kappa} = \begin{cases} 2\%, & 0 \leq \kappa < 1; \\ 51\%, & 1 \leq \kappa < 4; \\ 100\%, & \kappa \geq 4 \end{cases}$$

- ◇ increasing and right continuous
- ◇ discontinuous with possibly small discontinuity points
- ◇ practical importance

## Win-win-win situation

- Assume the mediator can charge 5% of their income

	Player 1	Player 2	mediator	Efficiency
$\kappa = 0$ (3, 1)	2.85	0.95	0.2	2%
$\kappa = 4$ (100, 100)	97	97	6	100%

$$\diamond 97 = 100 \times 95\% + 2, \quad 6 = (100 + 100) \times 5\% - 4$$

## Further remarks

- Mechanism 2 : taxation
  - ◇ Both rewarding and punishment, more power for the mediator
- $E_{\kappa} := \sup_{\pi \in \Pi_{\kappa}} E(\pi)$ 
  - ◇ Principal-agent problem with multiple agents
  - ◇ The problem is very challenging in continuous time models

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## The dynamic model

- Fix  $(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P})$ , and  $[0, T]$
- $N$ -players :  $\alpha = (\alpha^1, \dots, \alpha^N)$
- The model : for  $i = 1, \dots, N$ ,

$$dX_t^\alpha = b(t, X_t^\alpha, \alpha_t)dt + \sigma(t, X_t^\alpha)dB_t;$$

$$J_i(t, x, \alpha) := \mathbf{E}^{t,x} \left[ g_i(X_T^\alpha) + \int_t^T f_i(s, X_s^\alpha, \alpha_s^i) ds \right].$$

- ◇ Both  $B$  and  $X$  can be multidimensional
- ◇ The volatility control case :  $\sigma = \sigma(t, x, \alpha)$ , is more involved and is an ongoing work.
- ◇ For simplicity, in this talk we set  $d_B = d_X = 1$  and  $\sigma \equiv 1$ .

# The raw set value

- Definition :  $\alpha^* \in \mathcal{A}^N$  is a Nash equilibrium at  $(t, x)$ , denoted as  $\alpha^* \in NE(t, x)$ , if

$$J_i(t, x, \alpha^*) \geq J_i(t, x, \alpha^{*, -i}, \alpha^i), \quad \forall \alpha^i, \forall i.$$

- Raw set value :

$$\mathbb{V}_0(t, x) := \left\{ J(t, x, \alpha^*) : \alpha^* \in NE(t, x) \right\} \subset \mathbb{R}^N$$

# The set value

- For control problem :

$$V_0 := \sup_{\alpha} J(\alpha) = \lim_{\varepsilon \rightarrow 0} J(\alpha^\varepsilon), \quad \text{not} \quad V_0 := J(\alpha^*)$$

- Define  $NE_\varepsilon(t, x)$  in obvious sense and then **set value** :

$$\mathbb{V}(t, x) := \bigcap_{\varepsilon > 0} \mathbb{V}_\varepsilon(t, x) = \lim_{\varepsilon \rightarrow 0} \mathbb{V}_\varepsilon(t, x),$$

$$\mathbb{V}_\varepsilon(t, x) := \left\{ y : |y - J(\alpha^\varepsilon)| \leq \varepsilon \text{ for some } \alpha^\varepsilon \in NE_\varepsilon(t, x) \right\}.$$

- ◇  $\mathbb{V}$  is always closed
- ◇  $\mathbb{V} \supset \text{closure}(\mathbb{V}_0)$ , and the inclusion could be strict



## A few remarks

- For control problem ( $N = 1$ ) with value function  $v(t, x)$ ,  
 $\mathbb{V}(t, x) = \{v(t, x)\}$ , optimal control  $\alpha^*$  may not exist
- For two person zero-sum game ( $N = 2$ ) with game value  $v(t, x)$ ,  
 $\mathbb{V}(t, x) = \{(v(t, x), -v(t, x))\}$ , saddle point  $\alpha^*$  may not exist
- It is a lot easier to obtain  $\alpha^\varepsilon$  then to obtain  $\alpha^*$ . In that case

$$\mathbb{V} \neq \emptyset = \mathbb{V}_0$$

- Buckdahn-Cardaliaguet-Rainer(2004), Frei-dos Reis(2011), Lin (2012)

## The dynamic programming principle

- Recall, for control problem,

$$\begin{aligned} v(0, x) &= \sup_{\alpha_{[0,t]}} \mathbf{E}^{0,x} \left[ v(t, X_t^{\alpha_{[0,t]}}) + \int_0^t f(\dots) ds \right] \\ &= \mathbf{E}^{0,x} \left[ v(t, X_t^{\alpha_{[0,t]}^*}) + \int_0^t f(\dots) ds \right]. \end{aligned}$$

- Expecting DPP for raw set value :

$$\begin{aligned} \mathbb{V}_0(0, x) &= \left\{ J(t, \psi; 0, x, \alpha_{[0,t]}^*) : \text{all } \psi : \mathbb{R} \rightarrow \mathbb{R}^N, \alpha^* \in \mathcal{A}^N \right. \\ &\quad \left. \psi(x') \in \mathbb{V}_0(t, x'), \forall x' \in \mathbb{R}, \alpha_{[0,t]}^* \in NE(t, \psi; 0, x) \right\} \end{aligned}$$

- ◇ Subgame on  $[0, t]$  with terminal condition  $\psi$  :

$$J_i(t, \psi; 0, x, \alpha) := \mathbf{E}^{0,x} \left[ \psi_i(X_{t_2}^\alpha) + \int_0^t f(\dots) ds \right]$$

# The admissible controls

- The set value is extremely sensitive to admissible controls!
- Open loop controls  $\alpha_i = \alpha_i(t, B_{[0,t]})$  : DPP fails!
- State dependent closed loop controls  $\alpha_i = \alpha_i(t, X_t)$  : DPP fails!
- We have to allow for path dependent controls  $\alpha_i(t, X_{[0,t]})$ 
  - ◇ With  $\alpha \in \mathcal{A}_{path}^N$ , the (raw) set value is still state dependent, denoted as  $\mathbb{V}_{0,path}(t, x)$ .
  - ◇ We have truly path dependent  $\alpha^*$ , in particular,

$$\mathbb{V}_{0,state}(t, x) \neq \mathbb{V}_{0,path}(t, x).$$

# The dynamic programming principle

## Theorem (Feinstein-Rudloff-Z. 2020)

$$\mathbb{V}_0(0, x) = \left\{ J(t, \psi; 0, x, \vec{\alpha}_{[0,t]}^*) : \alpha^* \in NE_{path}(t, \psi; 0, x) \right. \\ \left. \psi : C([0, t]) \rightarrow \mathbb{R}^N, \quad \psi(x_{[0,t]}) \in \mathbb{V}_0(t, x_t) \right\}$$

- $J_i(t, \psi; 0, x, \vec{\alpha}) := E^{0,x} \left[ \psi(X_{[0,t]}^{\vec{\alpha}}) + \int_0^t f(\dots) ds \right]$ .
- The DPP holds when  $b, f, g$  are also path dependent
- The DPP holds for the set value  $\mathbb{V}$ , after obvious modification
- Abreu-Pearce-Stacchetti (1990), Sannikov (2007)

# Hamiltonian and PDE

- Introduce

$$h(t, x, z, a) := b(t, x, a)z + f(t, x, a) \in \mathbb{R}^N$$

- For control problem ( $N = 1$ ),

$$H(t, x, z) := \sup_a h(t, x, z, a),$$

$$\partial_t v + \frac{1}{2} \partial_{xx} v + H(t, x, \partial_x v) = 0.$$

- For two person zero-sum game ( $N = 2$ ) under Issacs condition :

$$H_1(t, x, z) := \inf_{a_1} \sup_{a_2} h_1(t, x, z, \vec{a}) = \sup_{a_2} \inf_{a_1} h_1(t, x, z, \vec{a}),$$

$$\partial_t v_1 + \frac{1}{2} \partial_{xx} v_1 + H_1(t, x, \partial_x v_1) = 0,$$

$$H_2 = -H_1, \quad v_2 = -v_1$$

## The set valued Hamiltonian $\mathbb{H}$

- Fix  $(t, x, z)$ , the mapping  $a \mapsto h(t, x, a)$  is a static game, then we may introduce set valued Hamiltonian  $\mathbb{H}(t, x, z) \subset \mathbb{R}^N$  naturally.
- For control problem ( $N = 1$ ),  $\mathbb{H}(t, x, z) = \{H(t, x, z)\}$
- For two person zero-sum game ( $N = 2$ ) :
  - ◇ under Issacs condition :  $\mathbb{H} = \{(H_1, -H_1)\}$
  - ◇ **without** Issacs condition :  $\mathbb{H}(t, x, z) = \emptyset$
  - ◇ Isaacs condition  $\iff \mathbb{H} \neq \emptyset$  ( $\mathbb{H}_0$  can be empty)

# Vector valued PDE approach for set values (Qiao-Z. 2024+)

- Assume  $\mathbb{H}(t, x, z) = \{H(t, x, z)\}$ , then  $\mathbb{V}(t, x) = \{v^H(t, x)\}$ ,

$$\partial_t v_i^H + \frac{1}{2} \partial_{xx} v_i^H + H_i(t, x, \partial_x \vec{v}^H) = 0, \quad i = 1, \dots, N.$$

◇ This covers the control problem and zero sum game problem under Issacs condition

- In the general case, roughly speaking (need the  $\varepsilon$ -approximations)

$$\mathbb{V}(t, x) = \left\{ v^H(t, x) : \text{all } H \text{ s.t. } H(t, x, z) \in \mathbb{H}(t, x, z), \forall(t, x, z) \right\}$$

- Hamadene-Lepeltier-Peng(1997), Bensoussan-Frehse(2000) :  
 Showed  $\supset$  in terms of  $\mathbb{V}_0$  and  $\mathbb{H}_0$

## Set valued PDE approach (Iseri-Z. (???)

$$\sup_{\eta \in \mathcal{T}_V, h \in \mathbb{H}(t, x, \partial_x V(t, x, y) + \eta)} n_V \cdot \left[ \partial_t V + \frac{1}{2} \partial_{xx} V + h - \text{tr} \left( \eta^\top \partial_x n_V \sigma + \frac{1}{2} \partial_y n_V \eta \eta^\top \right) n_V \right] (t, x, y) = 0$$

- $y$  is on the boundary of  $V(t, x)$
- The normal vector  $n_V(t, x, y)$  is part of the solution
- $\eta$  is on the tangent space :  $\eta \cdot n_V = 0$
- $\partial_t V, \partial_x V, \partial_{xx} V$  are appropriately defined set valued derivatives
- In terms of  $\mathbb{H}_0$ ,  $h$  part means

$$h(t, x, \partial_x V(t, x, y) + \eta, \bar{a}^*) \text{ over all equilibria } \bar{a}^*$$

- The key of the theory is a **set valued Itô formula** (Iseri-Z. 2023)



Thank you very much for your attention !