

# Set Values for Mean-Field Games

& Set Valued PDEs

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- Set Valued  $\vee$ : parameters  $(t, x, \mu, \dots)$   $\longrightarrow$  collections (instead of elements)

Set Valued Geometric Surface Evolutions (phase transitions, crystal formations, image processing)

Frameworks

Stochastic Viability & Target Problems

Dynamic Risk Measures

N-player Games

Multivariate Control Problems

Mean-field Games.

## Multivariate Control Problem

$$\mathbb{V}(t, x) = \{ \mathcal{T}(t, x, \alpha) : \forall \alpha \}$$

- DPP

i.e. time-consistency.

- PDE (I&FB)

## Mean-Field Games

$$\mathbb{V}(t, \mu) = \{ \mathcal{T}(t, \mu, \alpha) : \forall \alpha \text{ equilibrium} \}$$

- DPP

- Convergence  $(\mathbb{V}^N \rightarrow \mathbb{V})$

- PDE (Master Equation) ?

## Set Valued Calculus

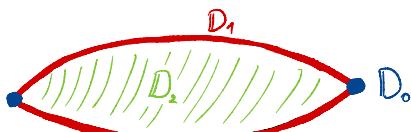
Defn  $D \in \mathcal{D}_2^n$  if it is closed and

- $D = \bigcup_{1 \leq k \leq m} D_k$ ,  $D_k$  is  $C^2$ -manifold with dimension  $k$ .
- $\exists n_1, \dots, n_{m-k} : D_k \rightarrow \mathbb{R}^m$  basis for normal space &  $\partial y^n(y)$  exists.

Defn  $T_D(y) = \left\{ y'(0) \mid y : \mathbb{R} \rightarrow D, y(0) = y \right\}$  tangent space

$$N_D(y) = (T_D(y))^\perp$$

normal space

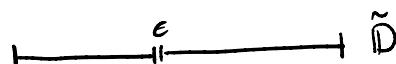
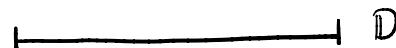


Given  $V(x) : \mathbb{R}^d \rightarrow \mathcal{D}_2^n$ :  $G_k \stackrel{\text{graph of}}{=} \{(x, y) : x \in \mathbb{R}^d, y \in V_k(x)\}$

Continuity  $V(x) : \mathbb{R}^d \rightarrow \mathcal{D}_2^n$  is continuous under all.

$$d(D, \tilde{D}) = \max_k d(D_k, \tilde{D}_k)$$

(Thausdorff)  $d(D, \tilde{D}) = \inf \{ \epsilon > 0 : \tilde{D} \subset O_\epsilon(D) \text{ & } D \subset O_\epsilon(\tilde{D}) \}$



$f : G_k \rightarrow \mathbb{R}$  is continuous

if  $(x_n, y_n) \in G_k \rightarrow (x, y) \in G_k$  then  $f(x_n, y_n) \rightarrow f(x, y)$

$$D = \bigcup_{1 \leq k \leq m} D_k$$

$D_k$  is  $C^2$ -manifold with dimension  $k$ .



$\nabla(x) : \mathbb{R}^d \rightarrow D_2^m$  is continuous under all.

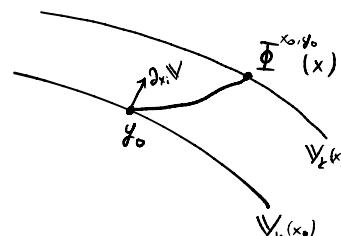
$$d(D, \tilde{D}) = \max_k d(D_k, \tilde{D}_k)$$

$f : G_k \rightarrow \mathbb{R}$  is continuous

if  $(x_n, y_n) \in G_k \rightarrow (x, y) \in G_k$  then  $f(x_n, y_n) \rightarrow f(x, y)$

Differentiability  $\nabla(x) : \mathbb{R}^d \rightarrow D_2^m$  is differentiable if  $\exists \Phi$  s.t.

$$\Phi^{x_0, y_0}(x_0) = y_0$$



$$\Phi^{x_0, y_0}(x) \in V_k(x) \text{ if } y_0 \in V_k(x_0)$$

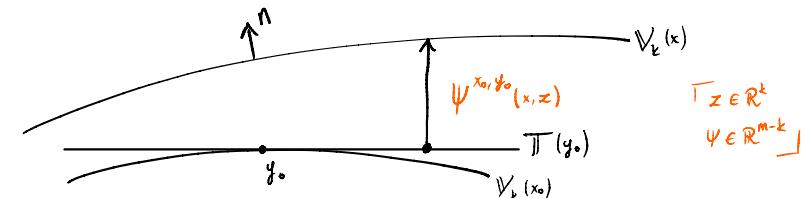
$\Phi^{x_0, y_0}(\cdot)$  is differentiable

$$\Phi'_i(x, y) = \partial_{x_i} \Phi^{x, y}(x) \text{ is continuous in } x \\ \text{Lipschitz in } y.$$

Moreover,  $\exists$  continuous basis  $n(x, y)$  for each  $k$ .

$$\bullet \quad \partial_{x_i} \nabla(x, y) \doteq (N \cdot \Phi'_i)(x, y) \text{ independent of } \Phi$$

Local Representation



$$\begin{aligned} \Gamma &\in \mathbb{R}^k \\ \Psi &\in \mathbb{R}^{m-k} \end{aligned}$$

$$- n \sim \partial_z \Psi \quad \& \quad \partial_y n \sim \partial_{zz} \Psi$$

Lemma If  $\nabla$  differentiable,

then  $\Psi$  is differentiable in  $x$  &  $\partial_z \Psi$  is continuous.

-  $D = \bigcup_{1 \leq k \leq m} D_k$

$D_k$  is  $C^2$ -manifold with dimension  $k$ .



-  $\mathbb{V}_k = \{ (x, y) \mid x \in \mathbb{R}^d, y \in \mathbb{V}_k(x) \}$  graph of  $\mathbb{V}_k$

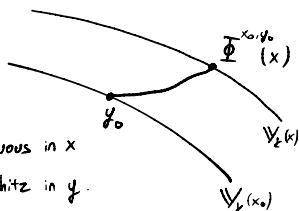
$\mathbb{V}(x) : \mathbb{R}^d \rightarrow \mathbb{D}_2^n$  is continuous under all.

$$d(D, \tilde{D}) = \max_k d(D_k, \tilde{D}_k)$$

$f : \mathbb{G}_k \rightarrow \mathbb{R}$  is continuous

if  $(x_n, y_n) \in \mathbb{G}_k \rightarrow (x, y) \in \mathbb{G}_k$  then  $f(x_n, y_n) \rightarrow f(x, y)$

### Differentiability



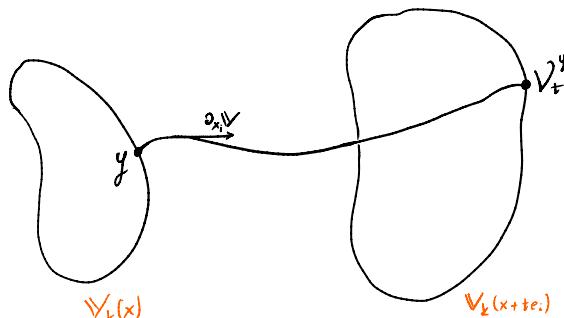
$\Phi'_i(x, y) = \partial_{x_i} \Phi^{x, y}(x)$  is continuous in  $x$   
Lipschitz in  $y$ .

$$\partial_{x_i} \mathbb{V}(x, y) = (N \cdot \Phi'_i)(x, y) \text{ independent of } \Phi$$

### Fundamental Theorem

$$V_t^i = y + \int_0^t \partial_{x_i} \mathbb{V}(x + se_i, V_s^i) ds$$

$$\mathbb{V}_k(x + te_i) = \{ V_t^y : y \in \mathbb{V}_k(x) \}$$



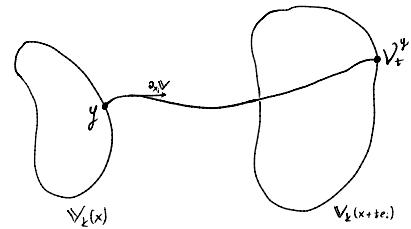
Defn  $\partial_{x_i} f(x, y) = \partial_s f(x + se_i, V_s^i)|_{s=0}$

Ex  $\partial_{x_i x_i} \mathbb{V}(x, y) = \partial_{x_i} (\partial_{x_i} \mathbb{V}(x, y))$

Ex  $\partial_{x^n}$  plays a role in Itô's Formula.

Remark  $\partial_{x_i x_i} \mathbb{V} \neq \partial_{x_j x_i} \mathbb{V}$

## Fundamental Theorem



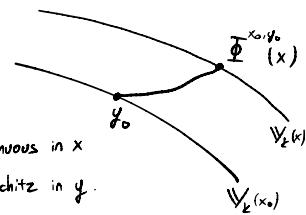
Defn  $\partial_{x_i} f(x, y) \doteq \partial_s f(x + se_i, V_s^i)|_{s=0}$

Ex  $\partial_{x_i} V(x, y) \doteq \partial_{x_i} (\partial_{x_i} V(x, y))$

Ex  $\partial_x n$  plays a role in Itô's Formula

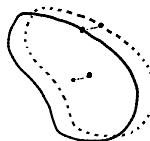
## Differentiability

$\bar{\Phi}_i'(x, y) \doteq \partial_{x_i} \bar{\Phi}^{x, y}(x)$  is continuous in  $x$   
Lipschitz in  $y$ .



$\partial_{x_i} V(x, y) \doteq (N \cdot \bar{\Phi}_i')(x, y)$  independent of  $\bar{\Phi}$

## Example 1 $V(x) = a(x) + 1$



$$\bar{\Phi}^{x_0, y_0}(x) = y_0 + a(x) - a(x_0)$$

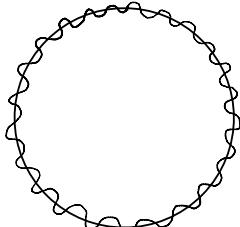
$$\Rightarrow \partial_x V(x, y) = N \cdot \partial_x a(x)$$

### Trivial Cases

- $D_0$
- Interior  $(\bullet)$

## Example 2 $V(x) = B(a(x), R(x))$ . $\partial_x V = n^\top \partial_x a + n \partial_x R$

## Example 3 $V(x) = \left\{ [1 + x \cos(1/x + m\theta)] (\cos\theta, \sin\theta) : \forall \theta \right\}$

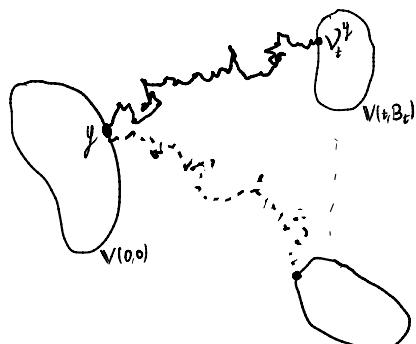


- continuous
- not differentiable
- $n$  is not continuous.

## Itô's Formula

$$V_t^y = y + \int_0^t \partial_t V + \frac{1}{2} \partial_{xx} V + K^e ds + \int_0^t \partial_x V + C dB_s$$

$$V_k(t, B_t) = \left\{ V_t^y : \forall y \in V_k(0, 0) \right\}$$



$$\begin{matrix} & K^e \\ \partial_x n & \diagdown \\ & \partial_y n \end{matrix}$$

MCP MFG

- DPP

- DPP

- Convergence

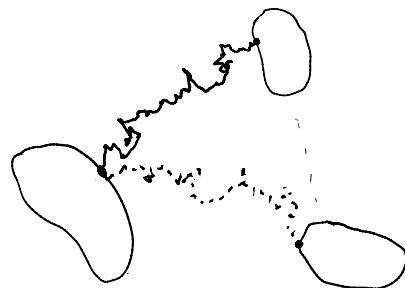
- PDE

- PDE?

### Ito's Formula

$$V_t^y = y + \int_0^t \partial_t V + \frac{1}{2} \partial_{xx} V + K^{\ell} ds + \int_0^t \partial_x V + \zeta dB_s$$

$$\mathbb{V}_k(t, B_t) = \{ V_t^y : \forall y \in \mathbb{V}(0, 0) \}$$



### Multivariate Control Setting

#### Dynamics

$$X_s^{t,x,\alpha} = x + \int_t^s b(r, X_r, \alpha_r) dr + \int_t^s \sigma(r, X_r, \alpha_r) dB_r$$

#### Value

$$Y_s^{t,x,\alpha} = g(X_T^{t,x,\alpha}) + \int_s^T F(r, X_r, \alpha_r, Y_r, Z_r) dr - \int_s^T Z_r^{t,x,\alpha} dB_r \in \mathbb{R}^m$$

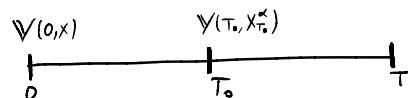
#### Set Value

$$\mathbb{V}(t, x) = \{ \mathcal{T}(t, x, \alpha) : \forall \alpha \}, \quad \mathcal{T}(t, x, \alpha) \doteq Y_t^{t,x,\alpha}$$

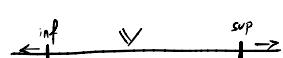
### DPP

$$\mathbb{V}(0, x) = \{ \mathcal{T}(0, x, \alpha; T_0, \phi) : \forall \alpha, \phi \in \mathbb{V}(T_0, X_{T_0}^\alpha) \}$$

$$\text{Recall: } V(0, x) = \sup_{\alpha} E \left[ V(T_0, X_{T_0}^\alpha) + \int_0^{T_0} F \dots \right]$$



m=1

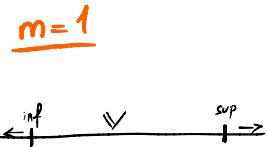


m > 1

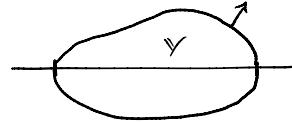


Set Value

$$\mathbb{V}(t, x) = \left\{ \mathcal{T}(t, x, \alpha) : \forall \alpha \right\}$$



$m > 1$



FFB

$$\sup_{\alpha, \ell} n \cdot \left[ \partial_t V + \frac{1}{2} \partial_{xx} V + K^\ell + F \right] (t, x, y, \alpha) = 0 \quad \forall (t, x, y) \in G$$

Theorem (i) Suppose  $V \in C^{1,2}$ . Then  $V$  solves FFB equation.

(ii) Suppose  $\tilde{V} \in C^{1,2}$  solves FFB and optimal  $I^*, \ell^*$  exists. Then  $V = \tilde{V}$ .

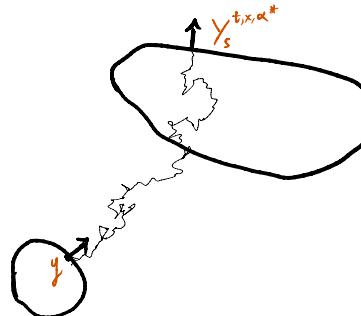
Moreover,  $y \in \mathbb{V}_b(t, x)$  and  $y = Y_t^{t, x, \alpha^*} \Rightarrow Y_s^{t, x, \alpha^*} \in \mathbb{V}_b(s, x_s^*)$

Theorem

(i) Suppose  $V \in C^{1,2}$ . Then  $V$  solves SFTB equation.

(ii) Suppose  $\tilde{V} \in C^{1,2}$  solves SFTB and optimal  $I^*, c^*$  exists. Then  $V = \tilde{V}$ .

Moreover,  $y \in V_b(t, x)$  and  $y = Y_i^{t, x, \alpha^*} \Rightarrow Y_s^{t, x, \alpha^*} \in V_b(s, x_s^*)$



## [Application] Moving Scalarization

$$\sup_{\alpha} Y_0^{0, x, \alpha} \cdot \lambda \xrightarrow{\text{optimal}} \alpha^* \text{ optimal}$$

~~$\alpha^*$  optimal~~

$$e. \sup_{\alpha} Y_t^{t, x_t^*, \alpha} \cdot \lambda \quad e. \sup_{\alpha} Y_t^{t, x_t^*, \alpha} \cdot n(t, x_t^*, y_t)$$

## Mean-Field Games



### Population

$$X_s^{t, \mu, \alpha} = X_t + \int_t^s b(r, X_r^{t, \mu, \alpha}, \mathcal{L}_{X_r^{t, \mu, \alpha}}, \alpha(r, X_r^{-}, \mathcal{L}_{X_r^{-}})) dr + B_{t,s}$$

### Player

$$X_s^{x, \tilde{\alpha}} = x + \int_t^s b(r, X_r^{x, \tilde{\alpha}}, \mathcal{L}_{X_r^{x, \tilde{\alpha}}}, \tilde{\alpha}(r, X_r^{-}, \mathcal{L}_{X_r^{-}})) dr + B_{t,s}$$

Nash Equilibrium

$$\mathcal{F}(\underbrace{t, \mu, \alpha^*}_{\text{population}}; \underbrace{\alpha^*, x}_{\text{player}}) \geq \mathcal{F}(t, \mu, \alpha^*; \alpha, x)$$

$$\mathbb{V}(t, \mu) \doteq \left\{ \mathcal{F}(t, \mu, \alpha^*; \alpha^*, \cdot) \mid \alpha^* \text{ equilibrium at } (t, \mu) \right\}.$$

Games are sensitive ..

---

MF G

population  $\alpha \in A$

player  $\tilde{\alpha} \in A$

$$\mathcal{T}(\alpha^*; \alpha^*) \geq \mathcal{T}(\alpha^*, \tilde{\alpha}) \quad \alpha^* \text{ equilibrium}$$

---

Relaxed

   $\gamma \in \mathcal{P}(A)$

   $\tilde{\gamma} \in \mathcal{P}(A)$

$$\mathcal{T}(\gamma^*; \gamma^*) \geq \mathcal{T}(\gamma^*, \tilde{\gamma}) \quad \gamma^* \text{ equilibrium}$$

---

Global

   $\Gamma \in \mathcal{P}(\alpha)$

   $\alpha \in A$

$$\mathcal{T}(\Gamma; \alpha) \geq \mathcal{T}(\Gamma, \tilde{\alpha}) \quad \begin{matrix} \Gamma \text{ equilibrium} \\ \alpha \in \text{supp } \Gamma \end{matrix}$$

•  $\nabla \neq \nabla^{\text{Relaxed}} = \nabla^{\text{Global}}$  •

Games are sensitive ..

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[Dynamics are state dependent]

but ..

state-dependent  $\propto (t, x_t, \mu_t)$

$$\bullet V^{\text{state}} \neq V^{\text{path}} \bullet$$

path-dependent  $\propto (t, x_{[0,t]}, \mu_{[0,t]})$

full-information  $\propto (t, x^1, \dots, x^N)$  — weak MFE

# Stability

$\epsilon$ -equilibrium

$$\int_{\mathbb{R}^d} \mathcal{T}(t, \mu, \alpha^*, \alpha^*, x) - \inf_{\alpha} \mathcal{T}(t, \mu, \alpha^*, \alpha, x) d\mu(x) \leq \epsilon$$

$$\mathbb{V}_\epsilon(t, \mu) \doteq \left\{ \varphi \text{ Lipschitz} : \right.$$

$$\left. \int_{\mathbb{R}^d} |\varphi - \mathcal{T}(t, \mu, \alpha^*, \alpha^*, \cdot)| d\mu(x) \leq \epsilon ; \quad \alpha^* \text{ } \epsilon\text{-equilibrium at } (t, \mu) \right\}$$

$$\mathbb{V}(t, \mu) \doteq \bigcap_{\epsilon > 0} \mathbb{V}_\epsilon(t, \mu)$$

Results. Diffusion: state: homogeneous discrete: state / path: homogeneous / heterogeneous

DPP  $\mathbb{V}(0, \mu) = \left\{ \mathcal{T} \left( t, \mu, \alpha^*, \alpha^*, \cdot; T_0, \varphi \right) : \alpha^* \text{ equilibrium}; \varphi \in \mathbb{V}(T_0, \mu_{T_0}^{\alpha^*}) \right\}$

- Under monotonicity (uniqueness) + I  $\hat{t} \Rightarrow$  Master Equation

### Convergence

$$\bigcap_{\epsilon > 0} \limsup_{N \rightarrow \infty} \mathbb{V}_\epsilon^{N, \text{hom}}(t, \vec{x}) = \mathbb{V}(t, \mu) = \bigcap_{\epsilon > 0} \mathbb{V}_\epsilon(t, \mu) \quad (\alpha_1 = \dots = \alpha_N)$$

$$\bigcap_{\epsilon > 0} \limsup_{N \rightarrow \infty} \mathbb{V}_\epsilon^{N, \text{het}}(t, \vec{x}) = \mathbb{V}^{\text{global}}(t, \mu) = \mathbb{V}^{\text{relax}}(t, \mu) \quad (\alpha_1, \dots, \alpha_N)$$

Thank you