

# Set Valued

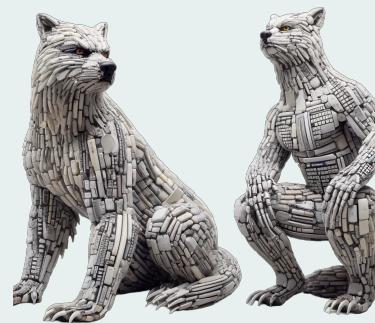
Melih İşeri

HJB

Joint work with Tianfeng Zhang.

# Equations

$$(t, x, \mu, \dots) \longrightarrow \dots \text{ or}$$



## Set Valued Frameworks

Geometric Surface Evolutions

Stochastic Viability & Target Problems

Multivariate Dynamic Risk Measures

[8] FEINSTEIN, Z.; RUDLOFF, B. (2015). *Multi-portfolio time consistency for set-valued convex and coherent risk measures*. Finance and Stochastics, **19**, 67-107.

[1] ARARAT, C.; MA, J.; WU, W. (2023). *Set-valued backward stochastic differential equations*. Annals of Applied Probability, **33**(5), 3418-3448.

N-player Games

[10] FEINSTEIN, Z.; RUDLOFF, B.; ZHANG, J. (2022). *Dynamic set values for nonzero sum games with multiple equilibria*. Mathematics of Operations Research, **47**, 616-642.

Mean-field Games

[13] İŞERİ, M.; ZHANG, J. (2021). *Set values for mean field games*. preprint, arXiv:2107.01661.

Multivariate Control Problems

## Set Valued Calculus

$$V(x) : \mathbb{R}^d \rightarrow \text{CR}^m$$

- closed, smooth boundary
- $V_b(x)$  is the boundary of  $V(x)$

- $\mathcal{G}_V \doteq \{(x, y) : x \in \mathbb{R}^d, y \in V_b(x)\}$

- $n_V(x, y) \doteq \mathcal{G}_V \rightarrow \mathbb{R}^m$  is the outward unit normal vector.

Intrinsic Derivative of  $f(x,y) : \mathbb{G}_V \rightarrow \mathbb{R}$ .

Defn  $\partial_x f(x,y) \doteq \lim_{\epsilon \rightarrow 0} \frac{f(x+\epsilon, \gamma^{x,y}(x+\epsilon)) - f(x,y)}{\epsilon}$

Ex  $\partial_x \gamma$  plays a role in Ito's Lemma.

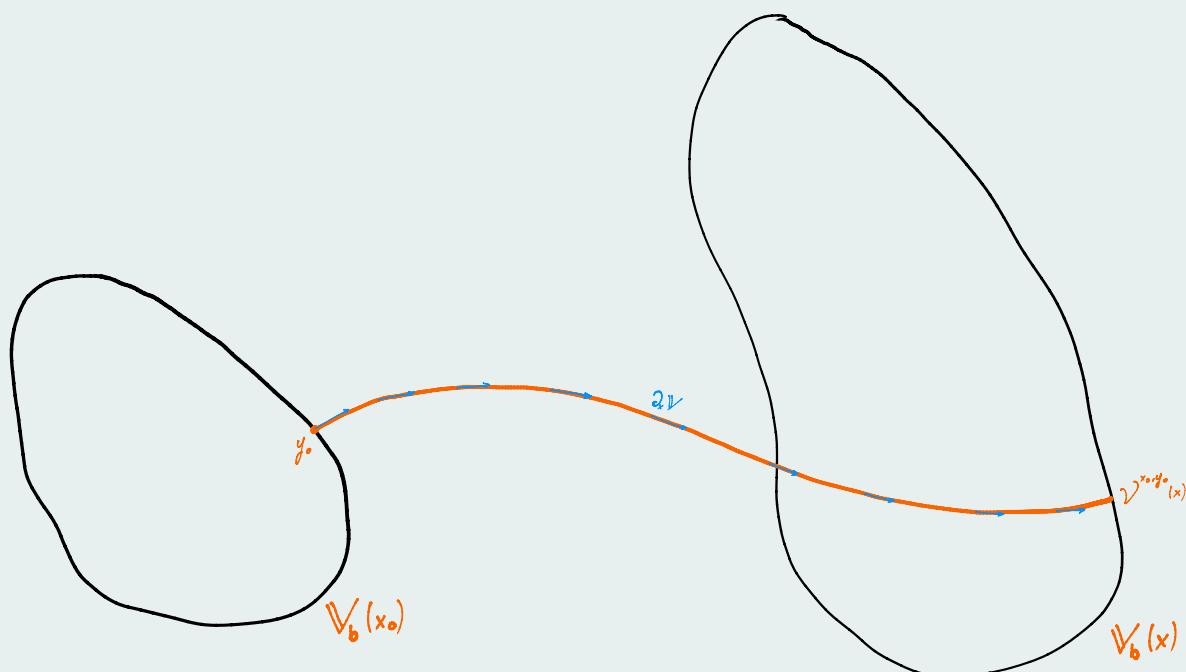
Defn  $\partial_x V(x,y) \doteq \partial_x(\gamma)$   $\Gamma_{f(x,y)=y}$

Defn  $\partial_{xx} V(x,y) \doteq \partial_x(\partial_x V(x,y))$

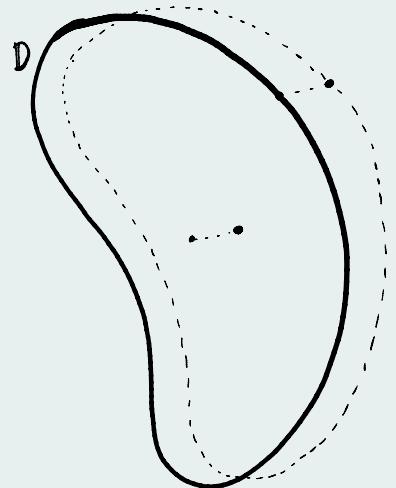
## Fundamental Theorem

$$\mathbb{V}_b(x) = \mathbb{V}_b(x_0) + \int_x^{x_0} \partial_x \mathbb{V}(\tilde{x}, \mathbb{V}_b(\tilde{x})) d\tilde{x}$$

$$\mathbb{V}_b(x) = \left\{ \mathcal{V}^{x_0, y_0}(x) : \forall y_0 \in \mathbb{V}_b(x_0) \right\} \quad \text{where} \quad \mathcal{V}^{x_0, y_0}(x) = x_0 + \int_{x_0}^x \partial_x \mathbb{V}(\tilde{x}, \mathcal{V}^{x_0, y_0}(\tilde{x})) d\tilde{x}$$



### Example 1



$$V(x) = \alpha(x) + D$$

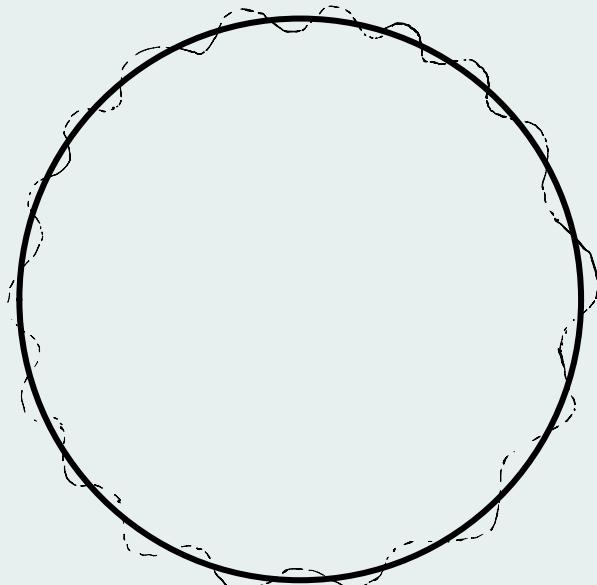
$$\partial_x V = n n^T \nabla_x \alpha(x)$$

### Example 2

$$V(x) = B(\alpha(x), R(x))$$

$$\partial_x V = n n^T \nabla_x \alpha + n \nabla_x R$$

### Example 3



$$V_b(x) = \left\{ [1 + x \cos(\frac{1}{x} + m\theta)] (\cos\theta, \sin\theta) : \forall \theta \right\}$$

- continuous
- NOT differentiable
- $n_V$  is not continuous.

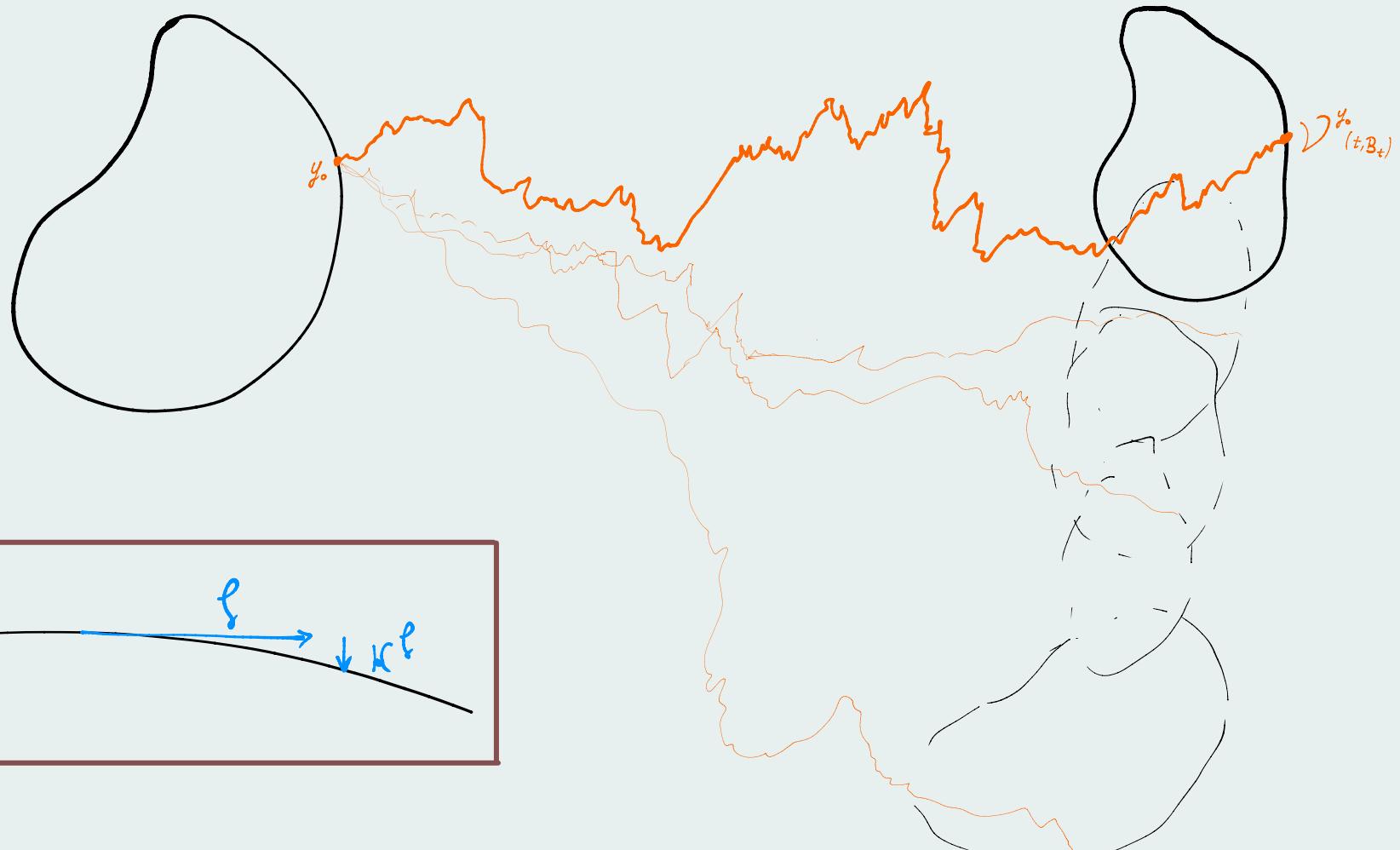
# Surface Evolution Equations

$$\partial_t \mathbb{V}(t, y) = h(t, y, n_{\mathbb{V}}, \partial_y n_{\mathbb{V}})$$

- [18] SETHIAN, J. A. (1985). *Curvature and the evolution of fronts*. Communications in Mathematical Physics, **101**, 487-499.
- [7] EVANS, L. C.; SPRUCK, J. (1991). *Motion of level sets by mean curvature. I*. Journal of Differential Geometry, **33**, 635-681.
- [4] BARLES, G.; SONER, H. M.; SOUGANIDIS, P. E. (1993). *Front propagation and phase field theory*. SIAM Journal on Control and Optimization, **31**(2), 439-469.
- [19] SONER, H. M. (1993). *Motion of a set by the curvature of its boundary*. Journal of Differential Equations, **101**, 313-372.
- [22] SONER, H. M.; TOUZI, N. (2003). *A stochastic representation for mean curvature type geometric flows*. The Annals of Probability, Vol. 31 No. 3, 1145-1165.
- [11] GIGA, Y. (2006). *Surface evolution equations: A level set approach*. Monographs in Mathematics, Birkhäuser Basel.

## Itô's Formula

$$V_b(t, B_t) = V_0(0, 0) + \int_0^t [\partial_t V + \frac{1}{2} \partial_{xx} V + \kappa^{\ell}] ds + \int_0^t [\partial_x V + \ell] dB_s$$



# Multivariate Control Problem

Dynamics

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

Value

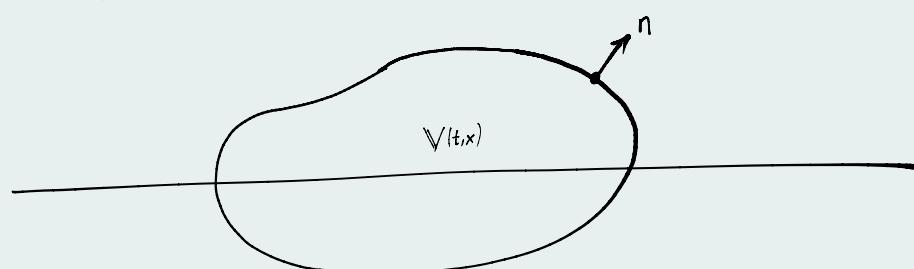
$$Y_s^{t,x,\alpha} = G(X_T^{t,x,\alpha}) + \int_s^T F(r, X_r^{t,x,\alpha}, \alpha_r, Y_r^{t,x,\alpha}, Z_r^{t,x,\alpha}) dr - \int_s^T Z_r^{t,x,\alpha} dB_r$$

Set-Value:  $\mathbb{V}(t,x) = \left\{ Y_t^{t,x,\alpha} : \forall \alpha \in \mathcal{A} \right\}$

In one dimension

$$\inf \left[ \begin{array}{c} \downarrow \\ | \\ \uparrow \end{array} \right] \mathbb{V}(t,x) \quad \sup \left[ \begin{array}{c} \uparrow \\ | \\ \downarrow \end{array} \right] \rightarrow$$

In higher dimensions



Motivation: Games, Portfolio of assets, Target Problems, Risk of many institutions.

Application: Time inconsistent problems

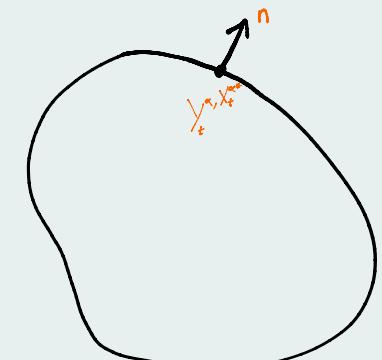
[15] KARNAM, C.; MA, J.; ZHANG, J. (2017). *Dynamic Approaches for Some Time Inconsistent Optimization Problems*. Annals of Applied Probability, 27, 3435-3477.

$$\text{Mean-Variance: } \sup_{\alpha} \mathbb{E}[X_T^{\alpha}] - \frac{\lambda}{2} (\mathbb{E}[|X_T^{\alpha}|^2] - \mathbb{E}[X_T^{\alpha}]^2) = \sup_{y \in \mathbb{V}(t,x)} \varphi(y),$$

$$Y^1 = X_T - \int Z^1 dB, \quad Y^2 = |X_T|^2 - \int Z^2 dB, \quad \varphi(y_1, y_2) = y_1 - \frac{\lambda}{2} y_1 + \frac{\lambda}{2} y_2^2$$

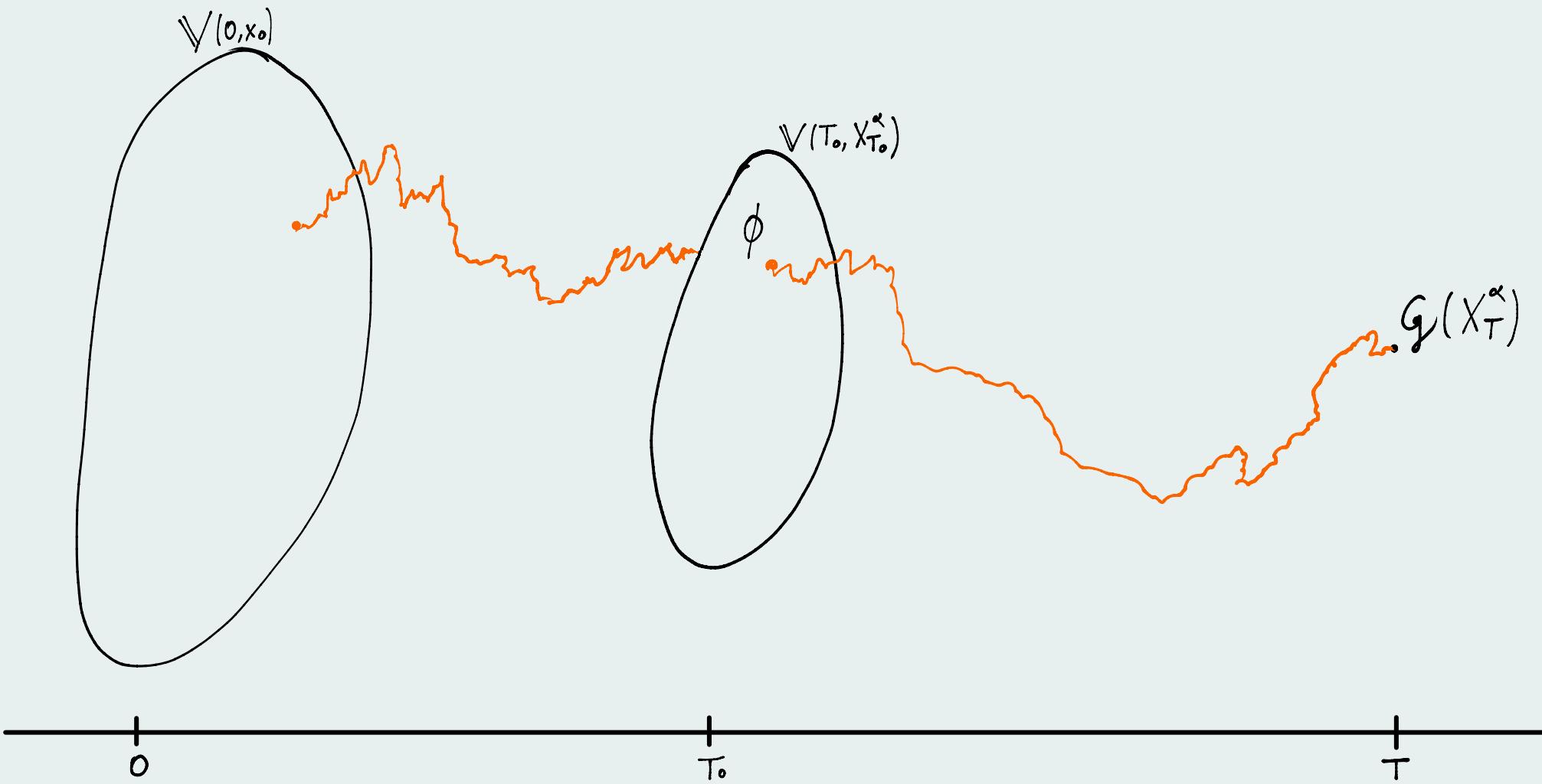
Moving Scalarization ( $\varphi(y) = \lambda \cdot y$ ):

$$\begin{aligned} & \sup_{\alpha} Y_0^{\alpha} \cdot \lambda \xrightarrow{\text{purple arrow}} \alpha^* \text{ optimal} \\ & \sup_{\alpha} Y_t^{\alpha}, X_t^{\alpha^*} \cdot \lambda \xleftarrow{\text{red arrow}} \\ & \sup_{\alpha} Y_t^{\alpha}, X_t^{\alpha^*} \cdot n(t, X_t^{\alpha^*}, Y_t^{\alpha^*}) \end{aligned}$$



# Time - Consistency

$$\mathbb{V}(0, x_0) = \left\{ Y_0^{T_0, \phi, \alpha} : \forall \alpha \in \mathcal{A}, \phi \in \mathbb{V}(T_0, X_{T_0}^\alpha) \right\}$$



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

$$\boxed{V(T, x) = g(x)}$$

### Main Theorem

(i) Suppose  $V \in C^{1,2}$ . Then  $V$  is a classical solution of HJB.

(ii) Suppose  $U \in C^{1,2}$  is a classical solution of HJB. Then  $V = U$ .

(ii') Suppose optimal arguments of Hamiltonian are  $I^*$  and  $g^*$ .

Then, for any  $y \in \mathbb{V}_b(t, x)$ ,  $\exists \alpha^* \text{ s.t. } y = y_t^{t, x, \alpha^*} \text{ and } y_s^{t, x, \alpha^*} \in \mathbb{V}(s, X_s^*)$

Moving Scalarization: Given  $\lambda$ , choose  $y^\lambda$  solving the  $\sup_y \lambda \cdot y$ .

By verification theorem, construct  $\alpha^*, X^*, Y^*$ , and since  $Y_t^* \in \mathbb{V}_b(t, X_t^*)$ ,  $\alpha^*$  is optimal for  $\Lambda(t, X_t^*, Y_t^*)$ ,  $\forall t$ .

Mean Variance: Introduce  $\Psi(y_1, y_2) = (y_1, y_2 - y_1^2)$  and  $\tilde{\mathbb{V}}(t, x) = \{\Psi(y) : \forall y \in \mathbb{V}(t, x)\}$

$\Rightarrow \sup_{y \in \mathbb{V}(t, x)} \varphi(y) = \sup_{\tilde{y} \in \tilde{\mathbb{V}}(t, x)} \tilde{y}_1 - \frac{\lambda}{2} \tilde{y}_2$  which is linear! Then, the dynamic mean-variance

$V_t = \text{ess} \sup_{\alpha} \left\{ \mathbb{E}[X_T^\alpha | \mathcal{F}_t] - \frac{\Lambda(t, X_{[0,t]}^*)}{2} \text{Var}(X_T^\alpha | \mathcal{F}_t) \right\}$  is time-consistent.

$$\alpha^* = -X_t^* + x_0 + \frac{1}{\lambda} e^T, \quad \Lambda(t, X_{[0,t]}) = \frac{\lambda e^{T-t}}{e^T - \lambda(x_t - x_0)}, \quad V_t = \frac{1}{2}(1+e^{t-T})X_t^* + \frac{1}{2}(1-e^{t-T})x_0 + \frac{e^T}{2\lambda}(1-e^{t-T})$$

thank  
you