

Stochastic Target Games with Controlled Loss

B. Bouchard

Ceremade - Univ. Paris-Dauphine, and, Crest - Ensae-ParisTech

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Joint work with L. Moreau (ETH-Zürich) and M. Nutz (Columbia)

Plan of the talk

- Problem formulation

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- Examples

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- Assumptions for this talk

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- Geometric dynamic programming

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- Geometric dynamic programming
- Application to the monotone case

Problem formulation and Motivations

Problem formulation

Provide a PDE characterization of the *viability* sets

$$\Lambda(t) := \{(z, p) : \exists u \in \mathcal{U} \text{ s. t. } \mathbb{E} \left[\ell(Z_{t,z}^{u[\vartheta], \vartheta}(T)) \right] \geq m \forall \vartheta \in \mathcal{V}\}$$

In which :

- \mathcal{V} is a set of admissible adverse controls
- \mathcal{U} is a set of admissible strategies
- $Z_{t,z}^{u[\vartheta], \vartheta}$ is an adapted \mathbb{R}^d -valued process s.t. $Z_{t,z}^{u[\vartheta], \vartheta}(t) = z$
- ℓ is a given loss/utility function
- m a threshold.

Application in finance

- $Z_{t,z}^{u[\vartheta],\vartheta} = (X_{t,x}^{u[\vartheta],\vartheta}, Y_{t,x,y}^{u[\vartheta],\vartheta})$ where
 - $X_{t,x}^{u[\vartheta],\vartheta}$ models financial assets or factors with dynamics depending on ϑ
 - $Y_{t,x,y}^{u[\vartheta],\vartheta}$ models a wealth process
 - ϑ is the control of the market : parameter uncertainty (e.g. volatility), adverse players, etc...
 - $u[\vartheta]$ is the financial strategy given the past observations of ϑ .
- Flexible enough to embed constraints, transaction costs, market impact, etc...

Examples

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- Expected loss control for $\ell(z) = -[y - g(x)]^-$
- Give sense to problems that would be degenerate under \mathbb{P} – a.s. constraints : B. and Dang (guaranteed VWAP pricing).

Examples

□ Constraint in probability :

$$\Lambda(t) := \{(z, p) : \exists u \in \mathcal{U} \text{ s.t. } \mathbb{P} \left[Z_{t,z}^{u[\vartheta], \vartheta}(T) \in \mathcal{O} \right] \geq m \forall \vartheta \in \mathcal{V}\}$$

for $\ell(z) = 1_{z \in \mathcal{O}}$, $m \in (0, 1)$.

\Rightarrow Quantile-hedging in finance for $\mathcal{O} := \{y \geq g(x)\}$.

Examples

□ Matching a P&L distribution = Multiple constraints in probability :

$$\{\exists \mathbf{u} \in \mathcal{U} \text{ s.t. } \mathbb{P} \left[\text{dist} \left(Z_{t,z}^{\mathbf{u}^{[\vartheta]}, \vartheta}(T), \mathcal{O} \right) \leq \gamma_i \right] \geq m_i \forall i \leq I, \forall \vartheta \in \mathcal{V}\}$$

(see B. and Thanh Nam)

Examples

□ Almost sure constraint :

$$\Lambda(t) := \{z : \exists u \in \mathcal{U} \text{ s.t. } Z_{t,z}^{u^{[\vartheta]}, \vartheta}(T) \in \mathcal{O} \mathbb{P} - \text{a.s. } \forall \vartheta \in \mathcal{V}\}$$

for $\ell(z) = \mathbf{1}_{z \in \mathcal{O}}$, $m = 1$.

\Rightarrow Super-hedging in finance for $\mathcal{O} := \{y \geq g(x)\}$.

Unfortunately not covered by our assumptions...

Setting for this talk

(see the paper for an abstract version)

Brownian diffusion setting

Brownian diffusion setting

- **State process** : $Z^{u[\vartheta], \vartheta}$ solves (μ and σ continuous, uniformly Lipschitz in space)

$$Z(s) = z + \int_t^s \mu(Z(r), u[\vartheta]_r, \vartheta_r) dr + \int_t^s \sigma(Z(r), u[\vartheta]_r, \vartheta_r) dW_r.$$

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- **Controls and strategies** :

Brownian diffusion setting

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- **Controls and strategies** :

- \mathcal{V} is the set of predictable processes with values in $V \subset \mathbb{R}^d$.
- \mathcal{U} is set of non-anticipating maps $u : \vartheta \in \mathcal{V} \mapsto \mathcal{U}$, i.e.

$$\{\vartheta_1 =_{(0,s]} \vartheta_2\} \subset \{u[\vartheta_1] =_{(0,s]} u[\vartheta_2]\} \quad \forall \vartheta_1, \vartheta_2 \in \mathcal{V}, s \leq T,$$

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where \mathcal{U} is the set of predictable processes with values in $U \subset \mathbb{R}^d$.

- The loss function ℓ has polynomial growth and is continuous.

The game problem

- **Worst expected loss** for a given strategy :

$$J(t, z, u) := \operatorname{ess\,inf}_{\vartheta \in \mathcal{V}} \mathbb{E} \left[\ell \left(Z_{t,z}^{u[\vartheta], \vartheta}(T) \right) \mid \mathcal{F}_t \right]$$

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- **The *viability sets*** are given by

$$\Lambda(t) := \{(z, p) : \exists u \in \mathfrak{U} \text{ s.t. } J(t, z, u) \geq p \text{ } \mathbb{P} - \text{a.s.}\}.$$

Compare with the formulation of games in Buckdahn and Li (2008).

Geometric dynamic programming principle

How are the properties
 $(z, m) \in \Lambda(t)$ and $(Z_{t,z}^{u[\vartheta], \vartheta}(\theta), ?) \in \Lambda(\theta)$
related ?

□ **First direction :** Take $(z, m) \in \Lambda(t)$ and $u \in \mathfrak{U}$ such that

$$\operatorname{ess\,inf}_{\vartheta \in \mathcal{V}} \mathbb{E} \left[\ell \left(Z_{t,z}^{u[\vartheta], \vartheta}(T) \right) \mid \mathcal{F}_t \right] \geq m$$

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To take care of the evolution of the worst case scenario conditional expectation, we introduce :

$$S_r^\vartheta := \operatorname{ess\,inf}_{\bar{\vartheta} \in \mathcal{V}} \mathbb{E} \left[\ell \left(Z_{t,z}^{u[\vartheta \oplus_r \bar{\vartheta}], \vartheta \oplus_r \bar{\vartheta}}(T) \right) \mid \mathcal{F}_r \right].$$

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Then

S^ϑ is a submartingale and $S_t^\vartheta \geq m$ for all $\vartheta \in \mathcal{V}$,

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Then

S^ϑ is a submartingale and $S_t^\vartheta \geq m$ for all $\vartheta \in \mathcal{V}$,

and we can find a martingale M^ϑ such that

$$S^\vartheta \geq M^\vartheta \text{ and } M_t^\vartheta = S_t^\vartheta \geq m.$$

□ We have for all stopping times θ (may depend on u and ϑ)

$$\operatorname{ess\,inf}_{\bar{\vartheta} \in \mathcal{V}} \mathbb{E} \left[\ell \left(Z_{t,z}^{u[\vartheta \oplus_{\theta} \bar{\vartheta}], \vartheta \oplus_{\theta} \bar{\vartheta}}(T) \right) \mid \mathcal{F}_{\theta} \right] = S_{\theta}^{\vartheta} \geq M_{\theta}^{\vartheta}.$$

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If the above is used uniformly in the strategies, θ takes finitely many values, we can find a covering formed of $B_i \ni (t_i, z_i)$, $i \geq 1$, such that

$$J(t_i, z_i; u) \geq M_{\theta}^{\vartheta} - \varepsilon \text{ on } A_i := \{(\theta, Z_{t,z}^{u[\vartheta], \vartheta}(\theta)) \in B_i\}.$$

□ We have for all stopping times θ (may depend on u and v)

$$\operatorname{ess\,inf}_{\bar{v} \in \mathcal{V}} \mathbb{E} \left[\ell \left(Z_{t,z}^{u[\bar{v} \oplus_{\theta} \bar{v}], \bar{v} \oplus_{\theta} \bar{v}}(T) \right) \mid \mathcal{F}_{\theta} \right] = S_{\theta}^{\bar{v}} \geq M_{\theta}^{\bar{v}}.$$

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Hence

$$K(t_i, z_i) \geq M_{\theta}^{\bar{v}} - \varepsilon \text{ on } A_i.$$

where

$$K(t_i, z_i) := \operatorname{ess\,sup}_{u \in \mathcal{U}} \operatorname{ess\,inf}_{\bar{v} \in \mathcal{V}} \mathbb{E} \left[\ell \left(Z_{t_i, z_i}^{u[\bar{v}], \bar{v}}(T) \right) \mid \mathcal{F}_{t_i} \right] \text{ is deterministic.}$$

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□ If K is lsc

$$K(\theta(\omega), Z_{t,z}^{u[\bar{v}], \bar{v}}(\theta)(\omega)) \geq K(t_i, z_i) - \varepsilon \geq M_{\theta}^{\bar{v}}(\omega) - 2\varepsilon \text{ on } A_i.$$

□ We have for all stopping times θ (may depend on u and ϑ)

$$\operatorname{ess\,inf}_{\vartheta \in \mathcal{V}} \mathbb{E} \left[\ell \left(Z_{t,z}^{u[\vartheta \oplus_{\theta} \bar{\vartheta}], \vartheta \oplus_{\theta} \bar{\vartheta}}(T) \right) \mid \mathcal{F}_{\theta} \right] = S_{\theta}^{\vartheta} \geq M_{\theta}^{\vartheta}.$$

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Hence

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□ If K is lsc

$$\underline{(Z_{t,z}^{u[\vartheta], \vartheta}(\theta(\omega)), M_{\theta}^{\vartheta}(\omega) - 3\varepsilon) \in \Lambda(\theta(\omega))}$$

□ To get rid of ε , and for non-regular cases (in terms of K and J), we work by approximation : One needs to **start from** $(z, m - \iota) \in \Lambda(t)$ and obtain

$$(Z_{t,z}^{u[\vartheta],\vartheta}(\theta), M_{\theta}^{\vartheta}) \in \bar{\Lambda}(\theta) \mathbb{P} - \text{a.s. } \forall \vartheta \in \mathcal{V}$$

where

$$\bar{\Lambda}(t) := \left\{ (z, m) : \begin{array}{l} \text{there exist } (t_n, z_n, m_n) \rightarrow (t, z, m) \\ \text{such that } (z_n, m_n) \in \Lambda(t_n) \text{ and } t_n \geq t \text{ for all } n \geq 1 \end{array} \right\}$$

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Remark : $M^{\vartheta} = M_{t,p}^{\alpha^{\vartheta}} := p + \int_t^{\cdot} \alpha_s^{\vartheta} dW_s$, with $\alpha^{\vartheta} \in \mathcal{A}$, the set of predictable processes such that the above is a martingale.

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Remark : The bounded variation part of S is useless : optimal adverse player control should turn S in a martingale.

□ **Reverse direction** : Assume that $\{\theta^\vartheta, \vartheta \in \mathcal{V}\}$ takes finitely many values and that

$$\underline{(Z_{t,z}^{u[\vartheta],\vartheta}(\theta^\vartheta), M_{t,m}^{\alpha^\vartheta}(\theta^\vartheta)) \in \Lambda(\theta^\vartheta) \mathbb{P} - \text{a.s. } \forall \vartheta \in \mathcal{V}.}$$

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Then

$$K(\theta^\vartheta, Z_{t,z}^{u[\vartheta],\vartheta}(\theta^\vartheta)) \geq M_{t,m}^{\alpha^\vartheta}(\theta^\vartheta).$$

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Then

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Play again with balls + concatenation of strategies (assuming smoothness) to obtain $\bar{u} \in \mathfrak{U}$ such that

$$\mathbb{E} \left[\ell \left(Z_{t,z}^{(u \oplus_{\theta^\vartheta} \bar{u})[\vartheta],\vartheta} (T) \right) | \mathcal{F}_\theta \right] \geq M_{t,m}^{\alpha^\vartheta}(\theta^\vartheta) - \varepsilon \mathbb{P} - \text{a.s. } \forall \vartheta \in \mathcal{V}.$$

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By taking expectation

$$\mathbb{E} \left[\ell \left(Z_{t,z}^{(u \oplus_{\theta^\vartheta} \bar{u})[\vartheta],\vartheta} (T) \right) \mid \mathcal{F}_t \right] \geq m - \varepsilon \mathbb{P} - \text{a.s. } \forall \vartheta \in \mathcal{V}$$

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so that

$$(z, m - \varepsilon) \in \Lambda(t).$$

□ To cover the general case by approximations, we need to start with

$$(Z_{t,z}^{u^{[\vartheta]},\vartheta}(\theta^\vartheta), M_{t,m}^{\alpha^\vartheta}(\theta^\vartheta)) \in \mathring{\Lambda}_\iota(\theta^\vartheta) \mathbb{P} - \text{a.s. } \forall \vartheta \in \mathcal{V},$$

where

$$\mathring{\Lambda}_\iota(t) := \{(z, p) : (t', z', p') \in B_\iota(t, z, p) \text{ implies } (z', p') \in \Lambda(t')\}.$$

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□ To concatenate while keeping the non-anticipative feature :

- **Martingale strategies** : α^ϑ replaced by $\alpha : \vartheta \in \mathcal{V} \mapsto \alpha[\vartheta] \in \mathcal{A}$ in a non-anticipating way (corresponding set \mathfrak{A})
- **Non-anticipating stopping times** : $\theta[\vartheta]$ (typically first exit time of $(Z^{u[\vartheta],\vartheta}, M^{\alpha[\vartheta]})$ from a ball.

The geometric dynamic programming principle

(GDP1) : If $(z, m - \iota) \in \Lambda(t)$ for some $\iota > 0$, then $\exists u \in \mathfrak{U}$ and $\{\alpha^\vartheta, \vartheta \in \mathcal{V}\} \subset \mathcal{A}$ such that

$$(Z_{t,z}^{u[\vartheta],\vartheta}(\theta), M_{t,m}^{\alpha^\vartheta}(\theta)) \in \bar{\Lambda}(\theta) \mathbb{P} - \text{a.s. } \forall \vartheta \in \mathcal{V}.$$

(GDP2) : If $(u, \alpha) \in \mathfrak{U} \times \mathfrak{A}$ and $\iota > 0$ are such that

$$(Z_{t,z}^{u[\vartheta],\vartheta}(\theta[\vartheta]), M_{t,m}^{\alpha[\vartheta]}(\theta[\vartheta])) \in \dot{\Lambda}_\iota(\theta[\vartheta]) \mathbb{P} - \text{a.s. } \forall \vartheta \in \mathcal{V}$$

for some family $(\theta[\vartheta], \vartheta \in \mathcal{V})$ of non-anticipating stopping times, then

$$(z, m - \varepsilon) \in \Lambda(t), \forall \varepsilon > 0.$$

Rem : Relaxed version of Soner and Touzi (2002) and B., Elie and Touzi (2009).

Application to the monotone case

□ **Monotone case** : $Z_{t,x,y}^{u[\vartheta],\vartheta} = (X_{t,x}^{u[\vartheta],\vartheta}, Y_{t,x,y}^{u[\vartheta],\vartheta})$ with values in $\mathbb{R}^d \times \mathbb{R}$ with $X_{t,x}^{u[\vartheta],\vartheta}$ independent of y and $Y_{t,x,y}^{u[\vartheta],\vartheta} \uparrow y$.

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□ **The value function is :**

$$\varpi(t, x, m) := \inf\{y : (x, y, m) \in \Lambda(t)\}.$$

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□ **Remark :**

□ **Monotone case** : $Z_{t,x,y}^{u[\vartheta],\vartheta} = (X_{t,x}^{u[\vartheta],\vartheta}, Y_{t,x,y}^{u[\vartheta],\vartheta})$ with values in $\mathbb{R}^d \times \mathbb{R}$ with $X_{t,x}^{u[\vartheta],\vartheta}$ independent of y and $Y_{t,x,y}^{u[\vartheta],\vartheta} \uparrow y$.

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(GDP1) : Let $\varphi \in C^0$ be such that $\arg \min(\varpi_* - \varphi) = (t, x, m)$. Assume that $y > \varpi(t, x, m - \iota)$ for some $\iota > 0$. Then, there exists $(u, a) \in \mathfrak{U} \times \mathfrak{A}$ that

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(GDP2) : Let $\varphi \in C^0$ be such that $\arg \max(\varpi^* - \varphi) = (t, x, m)$. Assume that $(u, \alpha) \in \mathfrak{U} \times \mathfrak{A}$ and $\eta > 0$ are such that

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Remark : In the spirit of the weak dynamic programming principle (B. and Touzi, and, B. and Nutz).

PDE characterization - “waving hands” version

- Assuming smoothness, existence of optimal strategies...
- $y = \varpi(t, x, m)$ implies
 $Y^{u[\vartheta], \vartheta}(t+) \geq \varpi(t+, X^{u[\vartheta], \vartheta}(t+), M^{a[\vartheta]}(t+))$ for all ϑ .

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Hence, for all ϑ ,

$$\begin{aligned}\mu_Y(x, y, u[\vartheta]_t, \vartheta_t) &\geq \mathcal{L}_{X, M}^{u[\vartheta]_t, \vartheta_t, a[\vartheta]_t} \varpi(t, x, m) \\ \sigma_Y(x, y, u[\vartheta]_t, \vartheta_t) &= \sigma_X(x, u[\vartheta]_t, \vartheta_t) D_x \varpi(t, x, p) \\ &\quad + a[\vartheta]_t D_m \varpi(t, x, m)\end{aligned}$$

with $y = \varpi(t, x, m)$

PDE characterization - “waving hands” version

□ Supersolution property

$$\inf_{v \in V} \sup_{(u, a) \in \mathcal{N}^v \varpi} \left(\mu_Y(\cdot, \varpi, u, v) - \mathcal{L}_{X, M}^{u, v, a} \varpi \right) \geq 0$$

where

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□ Subsolution property

$$\sup_{(u[\cdot], a[\cdot]) \in \mathcal{N}^{[\cdot]} \varpi} \inf_{v \in V} \left(\mu_Y(\cdot, \varpi, u[v], v) - \mathcal{L}_{X, M}^{u[v], v, a[v]} \varpi \right) \leq 0$$

where

$$\mathcal{N}^{[\cdot]} \varpi := \{\text{loc. Lip. } (u[\cdot], a[\cdot]) \text{ s.t. } (u[\cdot], a[\cdot]) \in \mathcal{N}^{\cdot} \varpi(\cdot)\}.$$

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