

Approximate Hedging and BSDEs with weak boundaries

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based on joint works with
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Motivation

- **Stock price** : (with large investor's strategy π)

$$\frac{dS_u^\pi}{S_u^\pi} = \mu(u, S_u^\pi, \pi_u) du + \sigma(u, S_u^\pi, \pi_u) dW_u$$

- **Wealth** process : (risk free interest rate $r = 0$)

$$dX_u^\pi = \pi_u \frac{dS_u^\pi}{S_u^\pi} = \pi_u [\mu(u, S_u^\pi, \pi_u) du + \sigma(u, S_u^\pi, \pi_u) dW_u]$$

- **Super Hedging** problem of claim $h(S_T^\pi)$:

$$\inf \{x \geq 0 : \exists \pi \in \mathcal{A} \text{ s.t. } X_T^{0,x,\pi} \geq h(S_T^\pi) \text{ } \mathbb{P} - \text{ps} \} .$$

\implies Prudential approach which leads to expensive prices

- **Quantile Hedging** of the claim $h(S_T^\pi)$: Given $p \in (0, 1)$, find

$$\inf \{x \geq 0 : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{P} [X_T^{t,x,\pi} \geq h(S_T^\pi)] \geq p \} .$$

How decreases the price when one accepts to keep some hedging risk ?

Agenda

- 1 Dual approach of Föllmer and Leukert
- 2 A stochastic target approach
- 3 Non Markovian BSDE representation

Explicit solution in a complete market

- Restriction to a **complete** market (super-replication \Leftrightarrow replication)
- Stock price under the (unique) Risk Neutral Measure \mathbb{Q} :

$$\frac{dS_u}{S_u} = \sigma(u, S_u) dW_u \quad (\text{independent on } \pi)$$

- **Wealth** process :

$$dX_u^\pi = \pi_u \sigma(u, S_u) dW_u$$

- Dual problem reformulation :

Maximize the probability of hedge for a given starting wealth x



$$\max_{\pi \in \mathcal{A}} \mathbb{P} [X_T^{0,x,\pi} \geq h(S_T)]$$

Föllmer and Leukert approach to quantile hedging

Maximize the probability of hedge for a given initial wealth x

$$\max_{\pi \in \mathcal{A}} \mathbb{P} [X_T^{0,x,\pi} \geq h(S_T)]$$

$$\max_{X \in \mathcal{L}_T^0} \mathbb{P} [X \geq h(S_T)] \quad \text{under the constraint} \quad \mathbb{E}^{\mathbb{Q}} [X] \leq x$$

$$A = \{X \geq h(S_T)\} \quad \quad \quad X = h(S_T)\mathbf{1}_A$$

$$\max_{A \in \mathcal{F}_T} \mathbb{P} [A] \quad \text{under the constraint} \quad \mathbb{E}^{\mathbb{Q}} [h(S_T)\mathbf{1}_A] \leq x$$

$$\max_{A \in \mathcal{F}_T} \mathbb{P} [A] \quad \text{under the constraint} \quad \mathbb{Q}^h[A] \leq \frac{x}{\mathbb{E}^{\mathbb{Q}} [h(S_T)]}, \quad d\mathbb{Q}^h := \frac{h(S_T)}{\mathbb{E}^{\mathbb{Q}} [h(S_T)]} d\mathbb{Q}$$

A interprets as the critical region while testing \mathbb{Q}^h against \mathbb{P} .

Neyman-Pearson lemma \implies optimal critical region $A^*(x)$

- Optimal strategy $\pi^*(x)$: the one which replicates $h(S_T)\mathbf{1}_{A^*(x)}$
- Quantile replication price : $x^*(p)$ such that $\mathbb{P}[A^*(x^*(p))] = p$

Solution in General Case

- **Pros :**
 - Explicit solution in some simple (but important) cases.
 - Generic solution of the form : $X_T^{0,x,\pi} = h(S_T) \mathbf{1}_A$
 - Similar structure in incomplete markets.
- **Cons :**
 - Resolution of the dual problem
 - Explicit solution not known in general (numerics)
 - In incomplete markets, the dual problem is a control problem : how to solve it ?
 - Relies heavily on the duality between super-hedgeable claims and risk neutral measures.

⇒ Alternative dynamic approach

The particular case of super-hedging

- The **super hedging** price at time 0

$$\inf \{x \geq 0 : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{P} [X_T^{0,x,\pi} \geq h(S_T^\pi)] = 1 \}$$

- **Dynamic version** of the super-hedging problem

$$v(t, s, 1) = \inf \{x \geq 0 : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{P} [X_T^{t,x,\pi} \geq h(S_T^{t,s,\pi})] = 1 \}$$

- **Dual approach** : $v(t, s; 1) = \sup_{\mathbb{Q}} \mathbb{E}^{\mathbb{Q}} [h(S_T^{t,s})]$

- **Direct approach** of Soner and Touzi :

- **(DP1)** : $x > v(t, s, 1) \Rightarrow \exists \pi \in \mathcal{A} \text{ s.t. for all stopping time } \tau \leq T$

$$X_\tau^{t,x,\pi} \geq v(\tau, S_\tau^{t,s,\pi}, 1)$$

- **(DP2)** : $x < v(t, s, 1) \Rightarrow \text{for all stopping time } \tau \leq T \text{ and } \pi \in \mathcal{A}$

$$\mathbb{P} [X_\tau^{t,x,\pi} > v(\tau, S_\tau^{t,s,\pi}, 1)] < 1$$

\Rightarrow **Allows to derive PDEs** associated to $v(\cdot, 1)$.

A stochastic target approach to quantile hedging

- The **quantile hedging** price at time 0

$$\inf \{x \geq 0 : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{P} [X_T^{0,x,\pi} \geq h(S_T^\pi)] \geq p \}$$

- Dynamic version** of the super-hedging problem

$$v(t, s, p) = \inf \{x \geq 0 : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{P} [X_T^{t,x,\pi} \geq h(S_T^{t,s,\pi})] \geq p \}$$

- Non consistent** dynamic problem
- Idea** : consider the "**probability of super-hedging**" as a **process** $(P_s)_{s \leq t \leq T}$
- This process must be a **martingale** and therefore of the form

$$P_s^{t,p,\alpha} = p + \int_t^s \alpha_u dW_u, \quad t \leq s \leq T, \quad \text{with } \alpha \in \mathbf{L}^2$$

- The quantile hedging price rewrites

$$v(t, s, p) = \inf \left\{ x \geq 0 : \exists \pi \in \mathcal{A} \text{ and } \alpha \in \mathbf{L}^2 \text{ s.t. } \mathbf{1}_{X_T^{t,x,\pi} \geq h(S_T^{t,s,\pi})} \geq P_T^{t,p,\alpha} \right\}$$

Dynamic programming for quantile replication

- **Dynamic** version of the quantile hedging price :

$$v(t, s, p) := \inf \left\{ x \in \mathbb{R}, \quad \exists (\pi, \alpha) \in \mathcal{A} \times \mathbf{L}^2 \quad \text{s.t.} \quad \mathbf{1}_{X_T^{t,x,\pi} \geq h(S_T^{t,s,\pi})} \geq P_T^{t,p,\alpha} \right\}$$

- Dynamic programming principle :

(DP1) : Starting with a wealth at time t **greater than** $v(t, s, p)$,
 one can at any time $\tau \geq t$ be able to (P_τ) -quantile replicate :

$$x > v(t, s, p) \Rightarrow \exists (\pi^*, \alpha^*) \quad \text{s.t.} \quad X_\tau^{t,x,\pi^*} \geq v(\tau, S_\tau^{t,s,\pi^*}, P_\tau^{t,p,\alpha^*}), \quad \forall \tau \in [t, T]$$

(DP2) : Starting with a wealth at time t **lower than** $v(t, s, p)$,
 it is impossible to quantile replicate :

$$x < v(t, s, p) \Rightarrow \forall (\pi, \alpha) \quad \mathbb{P} [X_\tau^{t,x,\pi} > v(\tau, S_\tau^{t,s,\pi}, P_\tau^{t,p,\alpha})] < 1, \quad \forall \tau \in [t, T]$$

Formal derivation of the Hamilton Jacobi Bellman equation

- Portfolio dynamics :
$$dX_r^\pi = \mu(r, S_r^\pi, \pi_r) \pi_r dr + \sigma(r, S_r^\pi, \pi_r) \pi_r dW_r$$

- Dynamics of $v(\cdot, S_r^\pi, P_r^\alpha)$:

$$dv(r, S_r^\pi, P_r^\alpha) = \left[v_t + \mu S_r^\pi v_x + \frac{\sigma^2 S_r^\pi}{2} v_{xx} + \frac{\alpha^2}{2} v_{pp} + 2\alpha\sigma S_r^\pi v_{xp} \right] (r, S_r^\pi, P_r^\alpha) dr + [\sigma S_r^\pi v_x + \alpha_r v_p] (r, S_r^\pi, P_r^\alpha) dW_r$$

- Take $x \sim v(t, s, p)$:

$$(DP1) \Rightarrow \exists(\pi^*, \alpha^*) \text{ s.t. } X_\tau^{t,x,\pi^*} \geq v(\tau, S_\tau^{t,s,\pi^*}, P_\tau^{t,p,\alpha^*}), \quad \forall \tau \in [t, T]$$

$$(DP2) \Rightarrow \forall(\pi, \alpha) \quad \mathbb{P}[X_\tau^{t,x,\pi} > v(\tau, S_\tau^{t,s,\pi}, P_\tau^{t,p,\alpha})] < 1, \quad \forall \tau \in [t, T]$$

- Formally, we deduce the following **HJB equation**

$$\sup_{(\alpha, \pi)} \mu\pi - \left[v_t + \mu S v_s + \frac{\sigma^2 S}{2} v_{ss} + \frac{\alpha^2}{2} v_{pp} + 2\alpha\sigma S v_{sp} \right] (t, s, p) = 0$$

under the constraint $\sigma\pi = [\sigma S v_s + \alpha v_p](t, s, p)$

Rigorous derivation

- PDE dynamics in the domain :

$$\sup_{(\alpha, \pi)} \mu\pi - \left[v_t + \mu s v_s + \frac{\sigma^2 s}{2} v_{ss} + \frac{\alpha^2}{2} v_{pp} + 2\alpha\sigma s v_{sp} \right] (t, s, p) = 0$$

under the constraint $\sigma\pi = [\sigma s v_s + \alpha v_p](t, s, p)$

- Main technical difficulty : the auxiliary control α is not bounded.
- The auxiliary control α is directly related to the primal control π .
- Boundary conditions :

$$\text{at } p = 0+ : v(t, s, 0) = 0$$

$$\text{at } p = 1- : v(t, s, 1) \text{ is the super-replication price}$$

$$\text{at } t = T- : v(T, s, p) = pg(s)$$

- Possible numerical approximation of the solution via PDE scheme

Explicit resolution in the Black Scholes model

- PDE in the **Black Scholes** model :

$$v_t + \frac{\sigma^2 s^2}{2} v_{ss} - \frac{\sigma^2 s^2}{2} \frac{|v_{sp}|^2}{v_{pp}} - \frac{\mu^2}{2\sigma^2} \frac{v_p^2}{v_{pp}} + \mu s \frac{v_p v_{sp}}{v_{pp}} = 0 \quad \text{with } v(T, s, p) = pg(s)$$

- Introduction of the **Fenchel-Legendre transform** $\tilde{v}(t, s, \cdot)$ of $v(t, s, \cdot)$:

$$\tilde{v}(t, s, y) := \sup_{p \in [0,1]} pq - v(t, s, p)$$

- The Fenchel Legendre transform \tilde{v} "solves" the following **linear PDE**

$$\tilde{v}_t + \frac{\sigma^2 s^2}{2} \tilde{v}_{ss} + \mu s q \tilde{v}_{sq} + \frac{\mu^2}{2\sigma^2} q^2 \tilde{v}_{qq} = 0 \quad \text{with } \tilde{v}(T, s, q) = (q - g(s))^+$$

- We deduce the **probabilistic representation** :

$$\tilde{v}(t, s, q) = \mathbb{E}[(Q_T^{t,q} - h(S_T^{t,s}))^+] \quad \text{with } Q^{t,q} := q + \int_t^{\cdot} \frac{\mu}{\sigma} Q_s^{t,q} dW_s$$

- We retrieve v by re-applying the **Fenchel transform**.

Extensions

- **On the Dynamics :**

$$S^\pi = s + \int_t^\cdot \mu(S_u^\pi, \pi_u) du + \int_t^\cdot \sigma(S_u^\pi, \pi_u) dW_u$$

$$X^\pi = x + \int_t^\cdot \rho(S_u^\pi, X_u^\pi, \pi_u) du + \int_t^\cdot \beta(S_u^\pi, X_u^\pi, \pi_u) dW_u$$

- **On the Problems :** Given $\ell: \mathbb{R}^d \times \mathbb{R} \rightarrow \mathbb{R}$ and $p \in \text{Im}(\ell)$,

$$v(t, s; p) := \inf \{x \in \mathbb{R} : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{E}[\ell(S_T^{t,s,\pi}, X_T^{t,x,\pi})] \geq p\} .$$

- **Possible range of applications**

$$\ell(s, x) = \mathbf{1}\{x \geq g(s)\} \quad \Rightarrow \quad \text{Quantile Hedging}$$

$$\ell(s, x) = U([x - g(s)]^+) \text{ with } U \nearrow \text{concave} \quad \Rightarrow \quad \text{Loss function}$$

$$\ell(s, x) = U(x - g(s)) \text{ with } U \nearrow \text{concave} \quad \Rightarrow \quad \text{Indifference pricing}$$

- **Dynamic programming** based on the reformulation

$$v(t, s; p) = \inf \{x \in \mathbb{R}_+ : \exists (\pi, \alpha) \in \mathcal{A} \times \mathbf{L}^2 \text{ s.t. } \ell(S_T^{t,s,\pi}, X_T^{t,x,\pi}) \geq P_T^{t,p,\alpha}\} .$$

(good) leads for extensions...

- **Utility maximization** under **quantile hedging type constraint** :
 \implies PDE characterization but no numerics (at that point)

- Combination of **several constraints** :

Given $\ell_1, \ell_2, \dots, \ell_m$ and $p_i \in \text{Im}(\ell_i)$ for $i \leq m$,

$$v(t, s; p) := \inf \{ x \in \mathbb{R} : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{E} [\ell_i (S_T^{t,s,\pi}, X_T^{t,x,\pi})] \geq p_i, \forall i \leq m \}$$

\implies leads to **high dimensional** PDE, impossible to solve numerically

- **Robust quantile hedging** under model uncertainty

Given a class of model $(\mathbb{P}^\lambda)_\lambda$, try to quantile hedge in any model

$$\inf \left\{ x \in \mathbb{R} : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{E}^{\mathbb{P}^\lambda} [\ell(\lambda, S_T^{t,s,\pi}, X_T^{t,x,\pi})] \geq p_\lambda, \quad \forall \lambda \right\} .$$

\implies consider **dynamic games**

- **One day ahead constraint** :

Given a **time delay** $\delta > 0$, try to find

$$\inf \{ x \in \mathbb{R} : \exists \pi \in \mathcal{A} \text{ s.t. } \mathbb{E}_s [\ell(X_{s+\delta}^{t,x,\pi})] \geq p, \quad \forall s \leq T \} .$$

\implies hard to get a dynamic programming principle

- Consideration of **non markovian** terminal claim ξ .
- In a complete market, the replication price identifies as the solution of the BSDE (with no driver)

$$Y_t = Y_T - \int_t^T Z_s dW_s, \quad 0 \leq t \leq T \quad \text{with } Y_T = \xi$$

\implies Y price process and Z investment strategy (up to the volatility)

- In case of imperfections (e.g. portfolio constraints), the **super-replication price** of ξ identifies to the **minimal solution** to the BSDE

$$Y_t = Y_T - \int_t^T Z_s dW_s + \int_t^T dL_s, \quad 0 \leq t \leq T \quad \text{with } Y_T \geq \xi$$

where L is an increasing process.

- For the **quantile replication price**, we expect

$$Y_T \geq \xi \quad \text{to be replaced by} \quad \mathbb{P}(Y_T \geq \xi) \geq p$$

- Hence, this formally leads to a (no driver) **new type of BSDE** of the form

$$Y_t = Y_T - \int_t^T Z_s dW_s + \int_t^T dL_s, \quad 0 \leq t \leq T \quad \text{with } \mathbb{P}(Y_T \geq \xi) \geq p$$

- More generally, for an **increasing loss function** ℓ , we get

$$dY_t = Z_t dW_t - dL_t, \quad \text{with } \mathbb{E}[\ell(Y_T - \xi)] \geq p$$

- For a **random increasing** function ψ , we look towards the **minimal solution** to the new type of **BSDE**

$$dY_t = -g(t, Y_t, Z_t) dt + Z_t dW_t - dL_t, \quad \text{with } \mathbb{E}[\psi(Y_T)] \geq p$$

- Constraint on the terminal condition distribution

\implies "BSDE with **weak terminal condition**"

- For a **random increasing** function ψ and Lipschitz driver g , we look towards the **minimal solution** to the BSDE

$$dY_t = -g(t, Y_t, Z_t)dt + Z_t dW_t - dL_t, \quad \text{with } \mathbb{E}[\psi(Y_T)] \geq p$$

- Introduction of a **supplementary control** $\alpha \in \mathbf{L}^2$ and $P_t^{p,\alpha} := p + \int_0^t \alpha_s dW_s$
- Set of all possible terminal conditions : $(\psi^{-1}(P_T^{p,\alpha}))_{\alpha \in \mathbf{L}^2}$

- We suppose for simplicity $\psi : [0, 1] \rightarrow [0, 1]$

- Let $(Y^\alpha, Z^\alpha)_{\alpha \in \mathbf{L}^2}$ be the set of solutions to the **classical BSDEs**

$$dY_t^\alpha = -g(t, Y_t^\alpha, Z_t^\alpha)dt + Z_t^\alpha dW_t, \quad \text{with } Y_T^\alpha = \psi^{-1}(P_T^{p,\alpha})$$

- At any time t , we can rewrite $Y_t^\alpha = \mathcal{E}_{t,T}^g [\psi^{-1}(P_T^{p,\alpha})]$

- For $\alpha \in \mathbf{L}^2$, $(Y^\alpha = \mathcal{E}_{\cdot, T}^g [\psi^{-1}(P_T^{p, \alpha})], Z^\alpha)$ solves the classical BSDE

$$dY_t^\alpha = -g(t, Y_t^\alpha, Z_t^\alpha)dt + Z_t^\alpha dW_t, \quad \text{with } Y_T^\alpha = \psi^{-1}(P_T^{p, \alpha})$$

- Any Y -component of a **super-solution** to the **BSDE with weak terminal condition** is of the form Y^α .
- For any path α , in order to pay the **cheapest price**, we define :

$$\bar{Y}_t^\alpha := \text{essinf} \left\{ \mathcal{E}_{t, T}^g \left[\psi^{-1}(P_T^{p, \alpha'}) \right], \alpha' \in \mathbf{L}^2 \text{ s.t. } \alpha' = \alpha \text{ on } [0, t] \right\}, \quad \forall t$$

- Obtention of **Dynamic Programming Principle** for the family $(\bar{Y}^\alpha)_\alpha$

$$\bar{Y}_t^\alpha = \text{essinf} \left\{ \mathcal{E}_{t, t'}^g \left[\bar{Y}_{t'}^{\alpha'} \right], \alpha' \in \mathbf{L}^2 \text{ s.t. } \alpha' = \alpha \text{ on } [0, t] \right\}, \quad 0 \leq t \leq t' \leq T.$$

- \bar{Y}^α is indistinguishable from a **ladlag** g -submartingale

- For $\alpha \in \mathbf{L}^2$, $(Y^\alpha = \mathcal{E}_{\cdot, T}^g [\psi^{-1}(P_T^{P, \alpha})], Z^\alpha)$ solves the classical BSDE

$$dY_t^\alpha = -g(t, Y_t^\alpha, Z_t^\alpha)dt + Z_t^\alpha dW_t, \quad \text{with } Y_T^\alpha = \psi^{-1}(P_T^{P, \alpha})$$

- For any path α , in order to pay the **cheapest price**, we define :

$$\bar{Y}_t^\alpha := \operatorname{essinf} \left\{ \mathcal{E}_{t, T}^g \left[\psi^{-1}(P_T^{P, \alpha'}) \right], \alpha' \in \mathbf{L}^2 \text{ s.t. } \alpha' = \alpha \text{ on } [0, t] \right\}, \quad \forall t$$

- Additional assumption** : $\psi^{-1}(\omega, \cdot)$ is continuous for \mathbb{P} -a.e ω

$\implies \bar{Y}^\alpha$ is indistinguishable from a **cadlag** g -submartingale

- Characterization of **the family** $(\bar{Y}^\alpha)_{\alpha \in \mathbf{L}^2}$ of solutions :

$$\text{Dynamics : } \bar{Y}^\alpha = \psi^{-1}(P_T^{P, \alpha}) + \int_{\cdot}^T g(s, \bar{Y}_s^\alpha, \bar{Z}_s^\alpha) ds - \int_{\cdot}^T \bar{Z}_s^\alpha dW_s + \bar{L}^\alpha - \bar{L}_T^\alpha \text{ on } [0, T]$$

$$\text{Minimality : } \bar{L}_{\tau_1}^\alpha = \operatorname{essinf} \left\{ E \left[\bar{L}_{\tau_2}^{\alpha'} | \mathcal{F}_{\tau_1} \right], \alpha' \in \mathbf{L}^2 \text{ s.t. } \alpha' = \alpha \text{ on } [0, \tau_1] \right\}, \quad \forall \tau_1 \leq \tau_2$$

$$\text{Futur indep. : } \alpha' = \alpha \text{ on } [0, \tau] \implies (\bar{Y}^{\alpha'}, \bar{Z}^{\alpha'}, \bar{L}^{\alpha'}) \mathbf{1}_{[0, \tau]} = (\bar{Y}^\alpha, \bar{Z}^\alpha, \bar{L}^\alpha) \mathbf{1}_{[0, \tau]}.$$

- Regularity at time t of $P_t^\alpha \mapsto \bar{Y}_t^\alpha$?

- Introduction of a modulus of continuity :

$$Err_t(\eta) := \text{ess sup} \{ |\mathcal{E}_{t,T}^g[\psi^{-1}(M)] - \mathcal{E}_{t,T}^g[\psi^{-1}(M')]|, M, M' \text{ s.t. } E_t[|M - M'|^2] \leq \eta \}$$

- For any $t < T$, we get

$$|\bar{Y}_t^\alpha - \bar{Y}_t^{\alpha'}| \leq Err_t(\Delta(P_t^\alpha, P_t^{\alpha'})) + Err_t(\Delta(P_t^{\alpha'}, P_t^\alpha)),$$

where

$$\Delta : (\mu_1, \mu_2) \mapsto \frac{\mu_2 - \mu_1}{\mu_2} \mathbf{1}_{\{\mu_1 < \mu_2\}} + \frac{\mu_1 - \mu_2}{1 - \mu_1} \mathbf{1}_{\{\mu_1 > \mu_2\}}$$

- Similar properties on $\{P_t^\alpha = 0\}$ or $\{P_t^\alpha = 1\}$.
- For a Lipschitz map ψ^{-1} , stability results on classical BSDEs

$$\implies \bar{Y}_t^\alpha \text{ is } L^2\text{-continuous with respect to } P_t^\alpha.$$

- Whenever $g(.,.)$ and ψ^{-1} are convex, there exists $\hat{\alpha}$ such that $\bar{Y}^{\hat{\alpha}} = Y^{\hat{\alpha}}$
 \implies a BSDE with weak terminal condition boils down to a **classical BSDE**
- For any $t < T$, the solution \bar{Y}_t^α is \mathcal{F}_t -convex with respect to P_t^α .
(need to consider the l.s.c. envelope of the solution)
- Probabilistic proof of the property.
- **"Facelift"**
 \implies if ψ deterministic, one can **replace ψ^{-1} by its convex envelope**
 \implies similar solutions on $[0, T)$
- In a Markovian framework, natural link with the previous **PDEs**.

- Suppose that g and ψ^{-1} are convex + technical conditions
- Introduce \tilde{g} the **Fenchel** transform of g w.r.t. (y, z) .
- Introduce $\tilde{\psi}^{-1}$ the **Fenchel** transform of ψ^{-1} w.r.t. p .
- Consider the following **dual control problem** :

$$\tilde{Y}_0(\ell) := \inf_{(\nu, \lambda) \in \text{Dom}(\tilde{g})} E \left[\int_0^T L_s^{\nu, \lambda} \tilde{g}(s, \nu_s, \lambda_s) ds + L_T^{\nu, \lambda} \tilde{\psi}^{-1}(\ell / L_T^{\nu, \lambda}) \right]$$

$$\text{where} \quad L_t^{\nu, \lambda} = 1 + \int_0^t L_s^{\nu, \lambda} (\nu_s ds + \lambda_s dW_s)$$

- We have the following correspondence

$$\bar{Y}_0(p) = \sup_{\ell > 0} (p\ell - \tilde{Y}_0(\ell)) \quad \text{and} \quad \tilde{Y}_0(\ell) = \sup_{p > 0} (p\ell - \bar{Y}_0(p))$$

- Standard explicit relation between the optimizers

- Particular case of **deterministic coefficients** and driver independent of z
- For $\alpha \in \mathbf{L}^2$ (with $\alpha > 0$), recall that (Y^α, Z^α) is solution to the classical BSDE with driver g and terminal condition $\psi^{-1}(P_T^\alpha)$

- Denoting $B^\alpha := \int_0^\cdot \alpha_s dW_s$; $\hat{Y}^\alpha := -Y^\alpha$ and $\hat{Z}^\alpha := -Z^\alpha/\alpha$, we get

$$\hat{Y}^\alpha = \psi^{-1}(p + B_T^\alpha) + \int_\cdot^T -g(s, -\hat{Y}_s^\alpha) ds - \int_\cdot^T \hat{Z}_s^\alpha dB_s^\alpha, \quad \mathbb{P} - \text{a.s.}$$

- B^α behaves under the **canonical meas.** \mathbb{P}^α as B under the pullback one \mathbb{P}^α

$\implies \hat{Y}^\alpha$ under \mathbb{P}^α looks like $\hat{Y}^{\mathbb{P}^\alpha}$ under \mathbb{P}^α where $(\hat{Y}^{\mathbb{P}^\alpha}, \hat{Z}^{\mathbb{P}^\alpha})$ solves

$$\hat{Y}^{\mathbb{P}^\alpha} = \psi^{-1}(p + B_T) + \int_\cdot^T -g(s, -\hat{Y}_s^{\mathbb{P}^\alpha}) ds - \int_\cdot^T \hat{Z}_s^{\mathbb{P}^\alpha} dB_s, \quad \mathbb{P}^\alpha - \text{a.s.}$$

- Therefore, we get : $-\tilde{Y}_0 = \text{ess sup}_\alpha \hat{Y}_0^\alpha = \text{ess sup}_\alpha \hat{Y}_0^{\mathbb{P}^\alpha}$.

\implies Link with **2BSDE** solution but **no aggregation** procedure.

- Consider a **time running constraint** on the distribution of Y :

$$\mathbb{E}[\psi(Y_t)] \geq 0, \quad 0 \leq t \leq T.$$

- For any date t , the reflection is related to the **law of Y_t**
- Consider the BSDE dynamics

$$Y_t = \xi + \int_t^T g(s, Y_s, Z_s) ds - \int_t^T Z_s \cdot dB_s + K_T - K_t$$

with the previous constraint and the analogous **new Skorokhod condition** :

$$\int_0^T \mathbb{E}[\psi(Y_t)] dK_t = 0.$$

- Dynamically non consistent problem** but we derive the well posed-ness of the BSDE

$$\begin{aligned}
 Y_t &= \xi + \int_t^T g(s, Y_s, Z_s) ds - \int_t^T Z_s \cdot dB_s + K_T - K_t \\
 \mathbb{E}[\psi(Y_t)] &\geq 0, \quad 0 \leq t \leq T \\
 \int_0^T \mathbb{E}[\psi(Y_t)] dK_t &= 0.
 \end{aligned}$$

- First observation : K must be deterministic or there is no continuous minimal solution.

The classical penalization procedure is a priori non monotonic.

- Existence and uniqueness of the solution under the bi-Lipschitz condition :

$$c_1|x - y| \mathbf{1} \leq |h(x) - h(y)| \leq c_2|x - y|$$

- Use of a fixed point argumentation.
- The Skorokhod condition implies the minimality of the solution
(at least when the driver does not depend on Y)

$$Y_t = \xi + \int_t^T g(s, Y_s, Z_s) ds - \int_t^T Z_s \cdot dB_s + K_T - K_t$$
$$\mathbb{E}[\psi(Y_t)] \geq 0, \quad 0 \leq t \leq T$$
$$\int_0^T \mathbb{E}[\psi(Y_t)] dK_t = 0.$$

- Can we approximate the solution of this **mean-field** reflected BSDE by a reflected BSDEs ?
- If $B = (B^1, \dots, B^N)$ are independent BM, can we solve the coupled system ?

$$Y_t^{i,N} = \xi + \int_t^T g(s, Y_s^{i,N}, Z_s^{i,N}) ds - \int_t^T Z_s^N \cdot dB_s + K_T^{i,N} - K_t^{i,N}$$

with $\frac{1}{N} \sum_{i=1}^N \psi(Y_t^{i,N}) \geq 0$

- What are the **asymptotics** when $N \rightarrow \infty$?

...which are on tracks :

- Addition of **jumps**
- Consideration of a constraint in *non linear expectation*
- Consideration of **weak reflections** in a dynamically consistent manner :

$$\mathbb{E}[\psi(Y_\tau)] \geq 0, \quad \text{for any stopping time } \tau \leq T.$$

... which should be reasonable :

- extension to a **quadratic** driver
- BSDE for **utility maximization** with quantile hedging constraint

... which seem more challenging :

- Case of **coupled FBSDE** for insider models
- BSDE for quantile hedging under **portfolio constraints**
- Consideration of **one day ahead constraints** : $\mathbb{E}_t[\psi(Y_{t+\delta})] \geq 0$.
- 2BSDE with weak terminal condition for **robust quantile hedging**
- **Numerics** for BSDE with weak terminal condition ?