

Generalized stochastic target problems for pricing and partial hedging under loss constraints - Application in optimal book liquidation

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Modeling and managing financial risks

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VWAP guaranteed contract

- ▶ Liquidation K stocks during $[0, T]$.
- ▶ Guarantee a better than δ basis point w.r.t market VWAP:

$$\text{Guaranteed VWAP} = \underbrace{(1 + \delta \cdot 10^{-4})}_{\gamma} \text{VWAP}_{\text{mkt}} .$$

- ▶ Brokerage fee: ask for a *premium* such that, up to a functional:

$$\text{premium} + \text{realized gain} \geq 0 \text{ a.s.} \quad (1)$$

Mathematical modeling

- ▶ VWAP: Volume Weighted Average Price
- ▶ Cumulative trading volume L_t : continuous real-valued, non decreasing adapted process and

$$L_0 = 0, L_T = K. \quad (2)$$

- ▶ Stock price $X^{L,1}(t)$
- ▶ Cumulative market volume: $\Theta(t) := \int_0^t \vartheta(s) ds$.

Execution turnover : $dY^L(t) = X^{L,1}(t)dL_t$

Market turnover : $dX^{L,2}(t) = X^{L,1}(t)d\Theta_t = X^{L,1}(t)\vartheta(t)dt$

$$Y^L(0) = X^{L,2}(0) = 0.$$

Mathematical modeling cont.

- ▶ Due to *price impact* effect: stock price is influenced by trading activity
- ▶ In case of linear impact

$$\frac{dX^{L,1}(t)}{X^{L,1}(t)} = \mu(t, X^{L,1}(t))dt + \sigma(t, X^{L,1}(t))dW_t - \beta(t, X^{L,1}(t))dL_t .$$

- ▶ Realized gain in cash

$$\left(\frac{Y^L(T)}{K} - \gamma \frac{X^{L,2}(T)}{\Theta(T)} \right) K .$$

VWAP benchmark

- ▶ VWAP: involve (price, proportion of volume) jointly
- ▶ S : stock price, $(v_t)_{t \in [0, T]}$: density of trading volume
- ▶ Define

$$p_t = \frac{\int_0^t v_u du}{\int_0^T v_u du} \text{ then } p_0 = 0, p_T = 1 \text{ and VWAP} = \int_0^T S_t dp_t .$$

- ▶ VWAP benchmark, heuristically, minimize the following quantity (p^X, p^M : broker and market trading curve)

$$\int_0^T S_t dp_t^X - \int_0^T S_t dp_t^M$$

\implies Follow the market “ \equiv ” minimize $\|p^X - p^M\|$.

Volume curve

- ▶ In reality, unable to determine (p_t^M) , $t \in [0, T]$ before T !
- ▶ One of possible proxies:

$$p_t^X \in [l_t^M, u_t^M] : \text{so-called } \textit{trading envelope}.$$

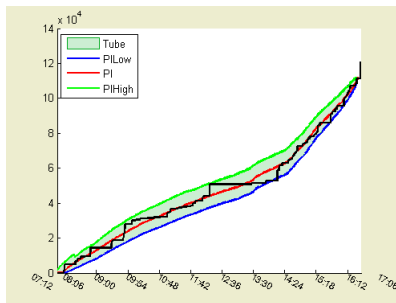


Figure: Trading envelopes

Mathematical modeling

- ▶ Introduce $X^{L,3}$:

$$X^{L,3}(t) = L_t - L_0 .$$

- ▶ Then require

$$X^{L,3}(s) \in [\underline{\Lambda}(s), \bar{\Lambda}(s)] \quad \text{for all } s \leq T , \quad (3)$$

where

$$\underline{\Lambda} < \bar{\Lambda} \text{ on } [0, T) \quad \text{and} \quad \underline{\Lambda}(T) = \bar{\Lambda}(T) = K .$$

- ▶ For future utilization, also require:

$$D\underline{\Lambda}, D\bar{\Lambda} \in (0, M] \text{ on } [0, T] \text{ for some } M > 0 .$$

Summarized problem

- ▶ For a control variable L , consider $(Y, X = (X^{L,1}, X^{L,2}, X^{L,3}), \Theta)$ whose dynamics are given as above.
- ▶ Given a function ℓ , finding y minimum such that (Equation (1))

$$\ell \left(y + \left(\frac{Y^L(T)}{K} - \gamma \frac{X^{L,2}(T)}{\Theta(T)} \right) K \right) \geq 0,$$

\implies Stochastic Target

Summarized problem

- ▶ For a control variable L , consider $(Y, X = (X^{L,1}, X^{L,2}, X^{L,3}), \Theta)$ whose dynamics are given as above.
- ▶ Given a function ℓ , finding y minimum such that (Equation (1))

$$\ell \left(y + \left(\frac{Y^L(T)}{K} - \gamma \frac{X^{L,2}(T)}{\Theta(T)} \right) K \right) \geq 0,$$

- ▶ and also

equation (2), (3) hold : $X^{L,3}(s) \in [\underline{\Lambda}(s), \bar{\Lambda}(s)]$ for all $s \leq T$.

⇒ Stochastic Target under State Constraints problem

Outline

Generalized stochastic target problem

- Abstract model

- Examples

- Geometric dynamic programming

- Informal PDE derivation

Liquidation problem

- Problem formulation

- PDE characterization

- Additional assumption and *a priori* estimates

- Comparison theorem

Conclusion

Problem formulation

- ▶ Given $\phi = (\nu, L) \in \mathcal{A} := \mathcal{U} \times \mathcal{L}$ the set of controls
- ▶ $Z^\phi = (X^\phi, Y^\phi) \in \mathbb{R}^d \times \mathbb{R}$ verifies:

$$dX^\phi = \mu_X(X^\phi, \nu_r)dr + \beta_X(X^\phi)dL + \sigma_X(X^\phi, \nu_r)dW_r$$

$$dY^\phi = \mu_Y(Z^\phi, \nu_r)dr + \beta_Y(Z^\phi)^\top dL + \sigma_Y(Z^\phi, \nu_r)^\top dW_r .$$

- ▶ Under Standing Assumption on constraint set:

$$\forall (t, x) : (x, y) \in O(t) , \quad y' \geq y \Rightarrow (x, y') \in O(t) .$$

- ▶ Then

$$V(t) := \left\{ (x, y) : Z_{t,x,y}^\phi(s) \in O(s) \quad \forall t \leq s \leq T \right\} .$$

is equivalent to

$$v(t, x) := \inf \{ y : (x, y) \in V(t) \} .$$

Super-hedging with cash delivery and proportional costs

- ▶ Let $d = 2$ and $\mu_X^i(x, u) = x^i \mu$, $\sigma_X^i(x, u) = x^i \sigma$, for $i \in \{1, 2\}$,

$$\beta_X^{21}(x) = -1, \quad \beta_X^{22}(x) = 1, \quad \beta_Y(x, y) = (1 - \lambda, -1 - \lambda).$$

- ▶ Then, $X^{L,1}$ follows a Black-Scholes dynamics, and

$$X_{t,x}^{2,L}(s) = x^2 + \int_t^s \frac{X_{t,x}^{2,L}(r)}{X_{t,x}^1(r)} dX_{t,x}^1(r) - \int_t^s dL_r^1 + \int_t^s dL_r^2$$

$$Y_{t,y}^L(s) = y + \int_t^s (1 - \lambda) dL_r^1 - \int_t^s (1 + \lambda) dL_r^2.$$

- ▶ Define $O(t) = \mathbf{1}_{t < T} \mathbb{R}_+^* \times \mathbb{R}^2 + \mathbf{1}_{t=T} \{(x, y) : \Lambda(y, x) \geq g(x)\}$
with $\Lambda(y, x) := y + (1 - \lambda)x^2$.

Loss function pricing

- ▶ Super-hedging criteria is too strict in markets with proportional costs
- ▶ Given a non-decreasing function ℓ , price at time t :

$$\hat{v}(t, x; p) := \inf \left\{ y : \mathbb{E} \left[\ell \left(\Lambda(Y^L(T), X^L(T)) - g(X^1(T)) \right) \right] \geq p \right\}$$

- ▶ Bouchard-Elie-Touzi (2009) shows that

$$\hat{v}(t, x; p) := \inf \left\{ y : G_{t,x,y}^L(T) \geq P_{t,p}^\nu(T), (\nu, L) \in \mathcal{U} \times \mathcal{L} \right\}$$

where

$$G_{t,x,y}^L(T) = \ell \left(\Lambda(Y_{t,y}^L(T), X_{t,x}^L(T)) - g(X_{t,x}^1(T)) \right) \in L^2$$

$$P_{t,p}^\nu := p + \int_t^\cdot \nu_s^1 dW_s^1 .$$

Geometric dynamic programming

- ▶ Firstly introduced by Soner and Touzi for super-hedging under Gamma constraints
- ▶ Extended to American type constraints: obstacle version of Bouchard-Vu (2010)

Theorem:

$$V(t) = \left\{ z : \exists \phi \in \mathcal{A} \text{ s.t. } Z_{t,z}^{\phi}(\theta \wedge \tau) \in O \bigoplus^{\tau, \theta} V \text{ for all } \theta, \tau \in \mathcal{T}_{[t, T]} \right\}$$

$$O \bigoplus^{\tau, \theta} V := O(\tau) 1_{\tau \leq \theta} + V(\theta) 1_{\tau > \theta} \text{ for } \theta, \tau \in \mathcal{T}_{[0, T]} .$$

Geometric dynamic programming

- ▶ Firstly introduced by Soner and Touzi for super-hedging under Gamma constraints
- ▶ Extended to American type constraints: obstacle version of Bouchard-Vu (2010)

Theorem: For all $\phi \in \mathcal{U} \times \mathcal{L}$ and $\theta \in \mathcal{T}_{[t, T]}$

1. GDP1: $y > v(t, x)$

$$Y_{t,x,y}^{\phi}(\theta) \geq v(\theta, X_{t,x}^{\phi}(\theta)) \text{ and } Z_{t,x,y}^{\phi} \in O \forall s \in [t, T].$$

2. GDP2: $y < v(t, x)$, then $\forall (\phi, \theta) \in \mathcal{A} \times \mathcal{T}_{[t, T]}$

$$\mathbb{P} \left[Y_{t,x,y}^{\phi}(\theta) > v(\theta, X_{t,x}^{\phi}(\theta)) \text{ and } Z_{t,x,y}^{\phi} \in O \forall s \in [t, T] \right] < 1.$$

Interior of the domain

In the case where $\beta_X = \beta_Y = 0$:

- ▶ $y = v(t, x)$, GDP implies that $\exists \nu \in \mathcal{U}$ such that:

$$\text{for } \phi = (\nu, 0) \text{ then } dY_{t,x,y}^\phi(t) \geq dv(t, X_{t,x}^\phi(t)). \quad (4)$$

- ▶ Formally:

$$dY^\phi(t) = \mu_Y^\phi(Z^\phi, \nu_t)dt + \sigma_Y(Z^\phi, \nu_t)dW_t$$

$$dv(t, X^\phi(t)) = \mathcal{L}_X^{\nu_t} v(t, X^\phi)dt + Dv(t, X^\phi(t))^\top \sigma_X(X^\phi, \nu_t)dW_t$$

- ▶ Inequality (4) suggests:

$$\mu_Y(Z^\phi, \nu_t) \geq \mathcal{L}_X^{\nu_t} v(t, X^\phi)$$

$$\text{and } \sigma_Y(Z^\phi, \nu_t) = \sigma_X(X^\phi, \nu_t)^\top Dv(t, X^\phi(t)),$$

Interior of the domain

In the case where $\beta_X, \beta_Y \neq 0$:

- ▶ $y = v(t, x)$, GDP implies that $\exists \nu \in \mathcal{U}$ such that:

$$\text{for } \phi = (\nu, L) \text{ then } dY_{t,x,y}^\phi(t) \geq dv(t, X_{t,x}^\phi(t)). \quad (4)$$

- ▶ Formally:

$$dY^\phi(t) = \dots + \beta_Y(Z^\phi)dL_t$$

$$dv(t, X^\phi(t)) = \dots + Dv(t, X^\phi(t))^\top \beta_X(X^\phi)dL_t$$

- ▶ Inequality (4) suggests:

$$\mu_Y(Z^\phi, \nu_t) \geq \mathcal{L}_X^{\nu_t} v(t, X^\phi)$$

$$\text{and } \sigma_Y(Z^\phi, \nu_t) = \sigma_X(X^\phi, \nu_t)^\top Dv(t, X^\phi(t)),$$

but in our case, also

$$\left(\beta_Y(Z^\phi)^\top - Dv(t, X^\phi(t))^\top \beta_X(X^\phi) \right) dL_t \geq 0.$$

Interior of the domain cont.

Define

$$F_\varepsilon^u := \sup \{ \mu_Y(x, y, u) - \mathcal{L}_X^u v, u \in N_\varepsilon v \}$$

$$G := \max \left\{ [\beta_Y(z)]^\top - Dv(t, x) \beta_X(x) \right] \ell, \ell \in \Delta_+ \right\}$$

where

$$N_\varepsilon v := \{ u \in U : |\sigma_Y(\cdot, v, u) - Dv \sigma_X(\cdot, u)| \leq \varepsilon \}$$

$$\Delta_+ := \mathbb{R}_+^d \cup \partial B_1(0).$$

then PDE characterization in the interior of the domain:

$$\max \{ F_0 v, G v \} = 0 \text{ on } (t, x, v(t, x)) \in \text{int}(D)$$

$$\text{where } D := \{ (t, x, y) : (x, y) \in O(t) \}.$$

On the boundaries of the domain

- ▶ Suppose $\exists \delta \in C^{1,2}$ such that $\delta = 0$ uniquely in ∂D , takes opposite sign inside and outside of D
- ▶ Then the state constraints require:

$$d\delta(Z_{t,z}^\phi(t)) \geq 0 \text{ if } (t, z) \in \partial D .$$

- ▶ Hence,

$$\mathcal{L}_Z^u \delta(t, x, y) \geq 0 \text{ and } D\delta(t, x, y)\sigma_Z(x, y, u) = 0 .$$

- ▶ Define F_0^{in} and G^{in} similarly as above, then PDE:

$$\max \left\{ F_0^{\text{in}} v, G^{\text{in}} v \right\} = 0 \text{ on } (t, x, v(t, x)) \in \partial D .$$

- ▶ Terminal condition + Relax all operators in $(\varepsilon, t, x, v, Dv, D^2v) \dots$

Mathematical modeling of the VWAP liquidation problem

Value function:

$$v(t, x, p) := \inf \left\{ y : \exists L \in \mathcal{L} \mid X_{t,x}^{3,L} \in [\underline{\Lambda}, \bar{\Lambda}] \text{ and } \mathbb{E} \left[\Psi(Z_{t,x,y}^L(T)) \right] \geq p \right\}$$

where $\Psi(x, y) = \ell(y - \gamma x^2)$.

Theorem:

$$v(t, x, p) := \inf \{ y \geq 0 : \mathcal{A}_{t,x,y,p} \neq \emptyset \},$$

where $\mathcal{A}_{t,x,y,p} := \{ (\nu, L) \in \mathcal{A} \mid (Z_{t,x,y}^L, P_{t,p}^\nu) \in V \text{ on } [t, T] \}$ with

$$V := \{ (x, y, p) : x^3 \in [\underline{\Lambda}, \bar{\Lambda}] \} \mathbf{1}_{[0, T]} \\ + \{ (x, y, p) : x^3 = K \text{ and } \ell(y - \gamma x^2) \geq p \} \mathbf{1}_{\{T\}},$$

and $P_{t,p}^\nu := p + \int_t^\cdot \nu_s dW_s$.

PDE characterization

Proposition: Under “good assumption”, v_* is a viscosity supersolution of

$$\max \{ F_0 \varphi, x^1 + x^1 \beta D_{x^1} \varphi - D_{x^3} \varphi \} = 0 .$$

And v^* is a subsolution of

$$\begin{aligned} \min \{ \varphi, \max \{ F_0 \varphi, x^1 + x^1 \beta D_{x^1} \varphi - D_{x^3} \varphi \} \} = 0 & \quad \text{if } \underline{\Lambda} < x^3 < \bar{\Lambda} \\ \min \{ \varphi, x^1 + x^1 \beta D_{x^1} \varphi - D_{x^3} \varphi \} = 0 & \quad \text{if } \underline{\Lambda} = x^3 \\ \min \{ \varphi, F_0 \varphi \} = 0 & \quad \text{if } x^3 = \bar{\Lambda} . \end{aligned}$$

Moreover,

$$v_*(T, x, p) = v^*(T, x, p) = \Psi^{-1}(x, p) \text{ for all } (x, p) \in [0, \infty)^2 \times \{K\} \times \mathbb{R} .$$

Additional assumption

- ▶ Operator F_0 defined as:

$$F_0\varphi := -\mathcal{L}_X\varphi - \frac{(x^1\sigma)^2}{2} \left(|D_{x^1}\varphi/D_p\varphi|^2 D_p^2\varphi - 2(D_{x^1}\varphi/D_p\varphi)D_{(x^1,p)}^2\varphi \right)$$

with

$$\mathcal{L}_X\varphi := \partial_t\varphi + x^1\mu D_{x^1}\varphi + x^1\vartheta D_{x^2}\varphi + \frac{1}{2}(x^1\sigma)^2 D_{x^1}^2\varphi.$$

- ▶ “Good assumption” on loss function ℓ

$$\begin{aligned} \exists \epsilon > 0 \text{ s.t. } \epsilon &\leq D^-\ell, \quad D^+\ell \leq \epsilon^{-1}, \\ \text{and } \lim_{r \rightarrow \infty} D^+\ell(r) &= \lim_{r \rightarrow \infty} D^-\ell(r) =: D\ell(\infty). \end{aligned}$$

- ▶ Also other conditions on boundaries $\underline{\Lambda}, \bar{\Lambda}$.

A priori estimates

- ▶ Proposition: $h \in (-(x^1 \wedge 1), 1)$

$$v(t, x, p) \geq v(t, x, p - \epsilon^{-1}|h|) + |h|$$

- ▶ Corollary:

v_* is a viscosity supersolution of

$$D_p \varphi - \epsilon = 0, (*)$$

and v^* is a viscosity subsolution of

$$-D_p \varphi + \epsilon = 0(**)$$

A priori estimates

- **Proposition:** $h \in (-(x^1 \wedge 1), 1)$, $\exists C$ depends only in x s.t.

$$v(t, x, p) \geq$$

$$v(t, x + he_1, p - C(x)|h|)$$

- **Corollary:**

v_* is a viscosity supersolution of

$$\min \{ (D_{x^1} \varphi - C(x) D_p \varphi) \mathbf{1}_{x^1 > 0}, -D_{x^1} \varphi + C(x) D_p \varphi \} = 0, (*)$$

and v^* is a viscosity subsolution of

$$\max \{ (D_{x^1} \varphi - C(x) D_p \varphi) \mathbf{1}_{x^1 > 0}, -D_{x^1} \varphi + C(x) D_p \varphi \} = 0 (**)$$

A priori estimates

- ▶ **Proposition:** $h \in (-(x^1 \wedge 1), 1)$, $\exists C$ depends only in x s.t.

$$v(t, x, p) \geq \max \left\{ v(t, x, p - \epsilon^{-1}|h|) + |h|, v(t, x + he_1, p - C(x)|h|) \right\}.$$

- ▶ **Corollary:**

v_* is a viscosity supersolution of

$$\min \left\{ D_p \varphi - \epsilon, (D_{x^1} \varphi - C(x) D_p \varphi) \mathbf{1}_{x^1 > 0}, -D_{x^1} \varphi + C(x) D_p \varphi \right\} = 0, (*)$$

and v^* is a viscosity subsolution of

$$\max \left\{ -D_p \varphi + \epsilon, (D_{x^1} \varphi - C(x) D_p \varphi) \mathbf{1}_{x^1 > 0}, -D_{x^1} \varphi + C(x) D_p \varphi \right\} = 0 (**)$$

Comparison principle and uniqueness

Assumption: $\exists \hat{x}^1 > 0$ s.t. $\mu(\cdot, \hat{x}^1) = \sigma(\cdot, \hat{x}^1) = 0$.

Theorem: U (resp. V) non-negative lower-semicontinuous supersolution (resp. upper-semicontinuous subsolution) and continuous in x^3 . Assume that

$$U(t, x, p) \geq V(t, x, p) \text{ if } t = T \text{ or } x^1 \in \{0, 2\hat{x}^1\},$$

and $\exists c_+ > 0, c_- \in \mathbb{R}$ such that

$$\limsup_{(t', x', p') \rightarrow (t, x, \infty)} V(t', x', p')/p' \leq c_+ \leq \liminf_{(t', y', p') \rightarrow (t, y, \infty)} U(t', y', p')/p',$$

$$\limsup_{(t', x', p') \rightarrow (t, x, -\infty)} V(t', x', p') \leq c_- \leq \liminf_{(t', y', p') \rightarrow (t, y, -\infty)} U(t', y', p').$$

If either U is a viscosity supersolution of (*) and continuous in p , or V is a viscosity subsolution of (**) and continuous in p , then

$$U \geq V.$$

Conclusion

- ▶ Propose generalized stochastic target problem:
 - ▶ controls in the form of bounded variation process
 - ▶ under state constraint.
- ▶ Suitable framework for:
 - ▶ pricing derivatives under loss constraint
 - ▶ models involving liquidity costs.
- ▶ Application in optimal liquidation: Pricing guaranteed VWAP contract with trading envelopes.
- ▶ Under “good assumptions”, comparison holds.
- ▶ Work on numerical resolution is in progress.

Thank you for your attention



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