Numerical approximation for a portfolio optimization problem under liquidity risk and costs.

presented by : M'HAMED GAIGI

Joint work with :

Vathana LY VATH, Mohamed MNIF and Salwa TOUMI







Motivation

Problem formulation

Oiscretized problem

Onvergence of the numerical scheme

Oumerical results

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- 2 Problem formulation
- 3 Discretized problem
- 4 Convergence of the numerical scheme
- 5 Numerical results

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Ly Vath V., Mnif M. and H. Pham.

A model of optimal portfolio selection under liquidity risk and price impact.

Finance and Stochastics, 11, 51-90, 2007.

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 - Control problem of portfolio optimization under liquidity risk and price impact.
 - The value function is the unique continuous viscosity solution of some HJB equation.
 - Numerical resolution of the impulse control problem under state constraints based on a probabilistic method.





- 2 Problem formulation
- 3 Discretized problem
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- We consider a continuous time financial market model. We denote by X_t the amount of money and by Y_t the number of shares in the stock held by the investor at time t. The price process of the risky asset is denoted by P_t .
- We model the investor's trades through an impulse control strategy $\alpha = (\tau_n, \zeta_n)_{n \ge 1}$, where the non-decreasing s.t. $\tau_1 \le \ldots \tau_n \le \ldots < T$ represent the intervention times and $(\zeta_n)_{n \ge 1}$ are \mathcal{F}_{τ_n} -measurable real valued r.v. and represent the number of stock trade at these times.

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Dynamics of Y

$$\begin{split} Y_s &= Y_{\tau_n}, \quad \tau_n \leq s < \tau_{n+1} \\ Y_{\tau_{n+1}} &= Y_{\tau_{n+1}^-} + \zeta_{n+1} \end{split}$$

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$$dP_s = P_s(bds + \sigma dW_s), \quad \tau_n \le s < \tau_{n+1}$$
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Dynamics of X

$$dX_s = rX_s ds \quad \tau_n \le s < \tau_{n+1}$$
$$X_{\tau_{n+1}} = X_{\tau_{n+1}^-} - \zeta_{n+1} e^{\lambda \zeta_{n+1}} P_{\tau_{n+1}^-} - k$$

State process

$$Z_s^{\alpha,t,z} = (X_s^{\alpha,t,x}, Y_s^{\alpha,t,y}, P_s^{\alpha,t,p}) \quad \forall s \in [t,T]$$

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The investor's net wealth

$$L(z) = \max[L_0(z), L_1(z)]\mathbf{1}_{y \ge 0} + L_0(z)\mathbf{1}_{y < 0}$$

where

$$L_0(z) = x + y p e^{-\lambda y} - k$$
, and $L_1(z) = x$.

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Solvency region

$$S = \left\{ z = (x, y, p) \in \mathbb{R} \times \mathbb{R} \times \mathbb{R}^*_+ : L(z) > 0 \right\},\$$

with

$$\partial S = \{z = (x, y, p) \in \mathbb{R} \times \mathbb{R} \times \mathbb{R}^*_+ : L(z) = 0\} \text{ and } \bar{S} = S \cup \partial S.$$

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Value Function

$$v(t,z) = \sup_{\alpha \in \mathcal{A}(t,z)} \mathbb{E}\left[e^{-r(T-t)}U_L(Z_T^{\alpha,t,z})\right], \quad (t,z) \in [0,T] \times \bar{S}.$$

with

$$U_L(z) = U(L(z)) = K(L(z))^{\gamma}, \quad \gamma \in]0,1[.$$

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HJB-QVI

$$\begin{split} \min\left[-\frac{\partial v}{\partial t} - \mathcal{L}v , v - \mathcal{H}v\right] &= 0 \quad sur \quad [0,T) \times S \\ \mathcal{L}\varphi &= rx\frac{\partial \varphi}{\partial x} + bp\frac{\partial \varphi}{\partial p} + \frac{1}{2}\sigma^2 p^2\frac{\partial^2 \varphi}{\partial p^2} - r\varphi, \\ \mathcal{H}\varphi(t,z) &= \sup_{\zeta \in \mathcal{C}(z)} \varphi(t,\Gamma(z,\zeta)), \quad (t,z) \in [0,T] \times \bar{S} \\ \Gamma(z,\zeta) &= (x - \zeta p e^{\lambda \zeta} - k, y + \zeta, p e^{\lambda \zeta}). \end{split}$$

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Theorem [Ly Vath, Mnif and Pham]

The value function v is continuous on $[0,T)\times S$ and is the unique (in $[0,T)\times S$) constrained viscosity solution to HJB-QVI satisfying the boundary and terminal conditions :

$$\lim_{\substack{(t',z') \to (t,z) \\ z' \in S}} v(t',z') = 0, \quad \forall (t,z) \in [0,T) \times D_0$$

$$\lim_{\substack{(t',z')\to(T,z)\\z'\in S}} v(t',z') = \max[U_L(z),\mathcal{H}U_L(z)], \quad \forall z\in \bar{S},$$

and

$$v(t,z)| \le K \left(1 + \left(x + \frac{p}{\lambda}\right)\right)^{\gamma}, \quad \forall (t,z) \in [0,T) \times S$$

where $D_0 = \{(0,0)\} \times \mathbb{R}^*_+$ and $K < \infty$.

$$\mathcal{G}_{\gamma}([0,T]\times\bar{S}) = \left\{ v: [0,T]\times\bar{S} \to \mathbb{R}; \sup_{[0,T]\times\bar{S}} \frac{|v(t,z)|}{(1+(x+\frac{p}{\lambda}))^{\gamma}} < \infty \right\}$$





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A classical way for a numerical approximation

Finite difference scheme

$$\frac{\partial \varphi}{\partial x}(x) \sim \frac{\varphi(x+\delta) - \varphi(x-\delta)}{2\delta}$$
$$\frac{\partial^2 \varphi}{\partial x^2}(x) \sim \frac{\varphi(x+\delta) - 2\phi(x) + \varphi(x-\delta)}{\delta^2}$$

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$$\varphi(t,z) = \max(\pounds_{\delta}\varphi(t,z); \sup_{\zeta \in \mathcal{C}_{\delta}(z)} \varphi(t,\Gamma(z,\zeta))$$

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Discretization scheme

$$S^{h}(t,z,\psi,\varphi) := \begin{cases} \min\left[\psi - \mathbb{E}[\varphi(t+h,Z_{t+h}^{0,t,z})], \psi - \mathcal{H}\varphi(t,z)\right]; t \in [0,T-h] \\ \min\left[\psi - \mathbb{E}[\varphi(T,Z_{T}^{0,t,z})], \psi - \mathcal{H}\varphi(t,z)\right]; t \in (T-h,T) \\ \min\left[\psi - U_{L}(z), \psi - \mathcal{H}U_{L}(z)\right]; t = T \end{cases}$$

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$$v^{h}(T,z) = \max\left[U_{L}(z), \mathcal{H}U_{L}(z)\right]$$
$$v^{h}(t_{i},z) = \max\left[\mathbb{E}[v^{h}(t_{i}+h, Z_{t_{i}+h}^{0,t_{i},z})], \mathcal{H}v^{h}(t_{i},z)\right],$$

where h = T/m and $i \in \{0, .., m - 1\}$.

Localized domain

$$\begin{split} \bar{S}_{loc} &= \bar{S} \cap \left([x_{min}, x_{max}] \times [y_{min}, y_{max}] \times [0, p_{max}] \right) \\ R &:= \min \Big(\mid x_{min} \mid, \mid x_{max} \mid, \mid y_{min} \mid, \mid y_{max} \mid, \mid p_{max} \mid \Big). \end{split}$$

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Space grid

$$\mathbb{Z}_{l} = \{ z = (x, y, p) \in \mathbb{X}_{l} \times \mathbb{Y}_{l} \times \mathbb{P}_{l}; z \in \bar{S}_{loc} \}$$

where \mathbb{X}_l is the uniform grid on $[x_{min}, x_{max}]$ of step $\frac{x_{max} - x_{min}}{l}$ and similarly for \mathbb{Y}_l and \mathbb{P}_l .

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where X_l is the uniform grid on $[x_{min}, x_{max}]$ of step $\frac{x_{max} - x_{min}}{l}$ and similarly for Y_l and \mathbb{P}_l .

Grid of the admissible controls

$$\mathcal{C}_{M,R}(z) = \{\zeta_i = \zeta_{min} + \frac{i}{M}(\zeta_{max} - \zeta_{min}); 0 \le i \le M/\Gamma(z,\zeta_i) \in \bar{S}_{loc}\}$$

where $\zeta_{min} < \zeta_{max} \in \mathbb{R}$ and $M \in \mathbb{N}^*$ are fixed constants.

14

Functional Quantization

$$\mathcal{E}^{N,R}[v^h(t, Z_t^{0,s,z})] := \sum_{i_1=1}^{N_1} \dots \sum_{i_{d(N)}=1}^{N_{d(N)}} \mathbb{P}_{i_1 \dots i_{d(N)}} v^h(t, Z_{N,R}^{0,s,z}(t)) \quad \forall \ s \le t$$

$$Z_{N,R}^{0,s,z}(t) := \left(x, y, proj_{[0,p_{max}]}(p \exp\left\{(b - \frac{\sigma^2}{2})(t - s) + \sigma W_{i_1..i_{d(N)}}^N(t - s)\right\})\right)$$
$$W_{i_1..i_{d(N)}}^N(t) = \sum_{n=1}^{d(N)} \sqrt{\lambda_n} x_{i_n} e_n(t) = \sum_{n=1}^{d(N)} \frac{\sqrt{2T}}{\pi(n - \frac{1}{2})} x_{i_n} \sin\left(\frac{\pi t}{T}(n - \frac{1}{2})\right)$$

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The optimal grid (x_{i_n}) and the associated weights $\mathbb{P}_{i_1..i_{d(N)}}$ are downloaded from the website : http ://www.quantize.maths-fi.com/downloads.

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H. Luschgya and G. Pages. Functional quantization of Gaussian processes. Journal of Functional Analysis, 196, 486–531, 2002.

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Discretization scheme

$$v^{h}(T,z) = \max\left[U_{L}(z), \sup_{\zeta \in \mathcal{C}_{M,R}(z)} U_{L}(\Gamma(z,\zeta))\right]$$
$$v^{h}(t_{i},z) = \max\left[\mathcal{E}^{N,R}[v^{h}(t_{i+1}, Z^{0,t_{i},z}_{t_{i+1}})], \mathcal{H}^{M,R}v^{h}(t_{i},z)\right]$$

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$$v_n(t,z) := \sup_{\alpha \in \mathcal{A}_n(t,z)} \mathbb{E}[U_L(Z_T)] \quad (t,z) \in [0,T] \times \bar{S}.$$

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$$v_n(t,z) := \sup_{\alpha \in \mathcal{A}_n(t,z)} \mathbb{E}[U_L(Z_T)] \quad (t,z) \in [0,T] \times \bar{S}.$$

Iterative scheme

We define the sequence $\varphi_n(t,z),$ solution of stopping time problems, as follows :

$$\varphi_{n+1}(t,z) = \sup_{\tau \in \mathcal{S}_{t,T}} \mathbb{E}[\mathcal{H}\varphi_n(\tau, Z_{\tau}^{0,t,z})],$$
$$\varphi_0(t,z) = v_0(t,z),$$

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Theorem

 v_n (hence φ_n) converges towards v when n goes to $+\infty$

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Approximation scheme

$$\begin{aligned} v_{n+1}^{h}(T,z) &= \max \left[U_{L}(z), \sup_{\zeta \in \mathcal{C}_{M,R}(z)} U_{L}(\Gamma(z,\zeta)) \right] \\ v_{n+1}^{h}(t_{i},z) &= \max \left[\mathcal{E}^{N,R}[v_{n+1}^{h}(t_{i+1}, Z_{t_{i+1}}^{0,t_{i},z})], \mathcal{H}^{M,R}v_{n}^{h}(t_{i},z) \right] \\ \text{for } i = 0, ..., m-1 \text{ ; } z = (x, y, p) \in \mathbb{Z}_{l} \text{ and starting from} \\ v_{0}^{h}(t,z) &= \mathcal{E}^{N,R}[U_{L}(Z_{T}^{0,t,z})] \end{aligned}$$

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Barles G. and P.E. Souganidis.

Convergence of approximation schemes for fully nonlinear second order equations.

Asymptotic analysis, 4, 271-283, 1991.

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Convergence

For all $(t,z) \in [0,T) \times S$ we have that

$$\lim_{\substack{(t',z')\to(t,z)\\(h,M,N,R)\to(0,+\infty)\\(t',z')\in\mathbb{T}_m\times\mathbb{Z}_l}} v^{h,M,N,R}(t',z') = v(t,z),$$

where $v^{h,R,N,M}$ is the solution of the discretized scheme and v is the solution of the HJB-QVI.

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Monotonicity

 $\forall \ h>0, \ (t,z)\in[0,T]\times\bar{S}, \ g\in\mathbb{R} \ \text{and} \ \varphi,\psi\in\mathcal{G}_{\gamma} \ \text{s.t.} \ \varphi\leq\psi \ \text{we have that}$

$$S^{h,R,N,M}(t,z,g,\varphi) \ge S^{h,R,N,M}(t,z,g,\psi)$$

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$$S^{h,R,N,M}(t,z,g,\varphi) \ge S^{h,R,N,M}(t,z,g,\psi)$$

Stability

For all h > 0, there exists a unique solution $v_n^{h,R,N,M} \in \mathcal{G}_{\gamma}([0,T] \times \overline{S})$ to the discretized scheme and the sequence $(v_n^{h,R,N,M})_h$ is uniformly bounded in $\mathcal{G}_{\gamma}([0,T] \times \overline{S})$ i.e. there exists $w \in \mathcal{G}_{\gamma}([0,T] \times \overline{S})$ s.t. $|v_n^{h,R,N,M}| \le |w|$ for all h > 0.

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Consistency

 $(i) \; \forall (t,z) \in [0,T) \times \bar{S}$ and Lipschitz function $\phi \in C^{1,2}([0,T) \times \bar{S})$ we have

$$\lim_{\substack{(h,t^{'},z^{'})\to(0,t,z)\\(M,N,R)\to+\infty}} \left\{ \frac{\phi(t^{'},z^{'}) - \mathcal{E}^{N,R}[\phi(t^{'}+h,Z^{0,t^{'},z^{'}})]}{h}, \phi(t^{'},z^{'}) - \mathcal{H}^{M,R}\phi(t^{'},z^{'}) \right\} \leq \min\left\{ \left(-\frac{\partial\phi}{\partial t} - \mathcal{L}\phi\right)(t,z), \left(\phi(t,z) - \mathcal{H}\phi(t,z)\right) \right\}$$

and

$$\lim_{\substack{(h,t^{'},z^{'})\to(0,t,z)\\(M,N,R)\to+\infty}} \left\{ \frac{\phi(t^{'},z^{'}) - \mathcal{E}^{N,R}[\phi(t^{'}+h,Z^{0,t^{'},z^{'}})]}{h}, \phi(t^{'},z^{'}) - \mathcal{H}^{M,R}\phi(t^{'},z^{'}) \right\} \\
\geq \min\left\{ \left(-\frac{\partial\phi}{\partial t} - \mathcal{L}\phi \right)(t,z), \left(\phi(t,z) - \mathcal{H}\phi(t,z)\right) \right\}$$

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Consistency (sequel)

 $(ii) \; \forall z \in \bar{S} \text{ and Lipschitz function } \phi \in C^{1,2}([0,T] \times \bar{S}) \text{ we have }$

$$\lim_{\substack{h,t^{'},z^{'})\to(0,T,z)\\(M,N,R)\to+\infty}} \min\left\{\phi(t^{'},z^{'}) - U_{L}(z^{'}), \left(\phi(t^{'},z^{'}) - \mathcal{H}^{M,R}U_{L}(z^{'})\right)\right\} \\
\leq \min\left\{\phi(T,z) - U_{L}(z), \left(\phi(T,z) - \mathcal{H}U_{L}(z)\right)\right\}$$

and

$$\lim_{\substack{(h,t',z')\to(0,T,z)\\(M,N,R)\to+\infty}} \min\left\{\phi(t',z') - U_L(z'), \left(\phi(t',z') - \mathcal{H}^{M,R}U_L(z')\right)\right\} \\
\geq \min\left\{\phi(T,z) - U_L(z), \left(\phi(T,z) - \mathcal{H}U_L(z)\right)\right\}$$

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Motivation

- Problem formulation
- 3 Discretized problem
- 4 Convergence of the numerical scheme
- 5 Numerical results

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Parameter	Value	Parameter	Value
Maturity	1 year	x_{min}	-100
λ	5.00E(-07)	x_{max}	200
γ	0.5	y_{min}	-4
σ	0.25	y_{max}	20
b	0.1	p_{min}	0
k	1	p_{max}	50
		l	20
		m	40
		M	100
		N	96
		$\bar{\varepsilon}$	10^{-3}

Table 1 : Parameters

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FIGURE : Value Function for fixed P

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FIGURE : The Optimal Policy sliced in XY

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FIGURE : The Optimal Policy sliced in XY for $\lambda = 5.00E(-03)$

Image: A image: A

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Convergence in N



FIGURE : Relative error of the value function computed when N = 96Vs. N = 200.

Problem formulation Convergence of the numerical scheme Numerical results

Convergence in M



FIGURE : Relative error of the value function when M = 200 Vs. M = 250. → M'hamed GAIGI AMaMeF and Banach Center Conference, Juin 2013

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Convergence in R

Some values of the value function for two different values R_1 and R_2 of R where R_1 is chosen as in Table 1 and we choose R_2 as follows :

$$\begin{aligned} R_2 &= \min \left(\mid x_{min} = -257.90 \mid, \mid x_{max} = 342.10 \mid, \mid y_{min} = -16.63 \mid, \\ \mid y_{max} = 31.36 \mid, \mid p_{max} = 100 \mid \right) \end{aligned}$$

R	R_1	R_2
$v(t,z_1)$	20.0038	19.9966
$v(t, z_2)$	24.5429	24.5340

Table 2 : Values of the value function for different values of R and z.

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Thank you for your attention.