

Mathematical Institute

Optimal Order Placement with Random Measures

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Oxford Mathematics Latency



- Latency is the time delay between an exchange streaming market data to a trader, the trader processing information and deciding to trade, and the exchange receiving the order from the trader.
- Some facts:
 - No market participant has zero latency!
 - Latency is random.
- Main consequences for liquidity
 - takers: marketable orders are not always executed/filled at the observed price,
 - makers: stale limit orders are picked off.





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Figure: Misses, latency, and discretion.

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Is there an optimal trade-off between costs and misses?



Outline:

- The model (fairly general)
- FBSDE arising from the vanishing Gâteaux derivative
- Global minimum
- Future work



Let $\mathcal{N} = \{(T_n, Z_n)\}_{(n \ge 1)}$ be a marked point process.

- (T_n) are the times at which the client trades
- (Z_n) are the changes in price due to latency

We associate the random measure p to the marked point process \mathcal{N}

$$p(\omega, [0, t], A) = \sum_{n} \mathbb{1}_{\{T_n(\omega) \leq t\}} \mathbb{1}_{\{Z_n(\omega) \in A\}}.$$

We assume the compensator is

$$\tilde{p}(\mathrm{d} z, \mathrm{d} t) = \phi_t(\mathrm{d} z) \,\lambda_t \mathrm{d} t \,.$$



For a given change in prices z, and discretion $\boldsymbol{\delta}$

The cost is

$$z F(\delta - z) = \begin{cases} z, & \text{if } z \le \delta & \text{trade filled} \\ 0, & \text{if } z > \delta & \text{trade missed} \end{cases}$$

where, $F(x) \coloneqq \mathbb{1}_{x \ge 0}$.

A miss trade is

$$D(z) = G(\delta - z) = \begin{cases} 0, & \text{if } z \le \delta \\ 1, & \text{if } z > \delta \end{cases} \text{ trade filled}$$

which is $1 - F(\delta - z)$.



• The running **cost** of strategy δ is

$$C_t^{\delta} = \int_0^t \int_{\mathbb{R}} z F(\delta_s - z) p(\mathrm{d}z, \mathrm{d}s).$$

The running number of missed trades is

$$D_t^{\delta} = \int_0^t \int_{\mathbb{R}} \left(1 - F\left(\delta_s - z\right)\right) \, p(\mathrm{d}z, \mathrm{d}s) \, .$$



We study the functional $J: \mathcal{A} \mapsto \mathbb{R}$

$$J(\delta) = \mathbb{E}\left[C_T^{\delta} + \boldsymbol{\alpha} D_T^{\delta} + \boldsymbol{\gamma} \left(D_T^{\delta}\right)^2\right],$$

where $\alpha, \gamma \ge 0$ and set of admissible strategies

$$\mathcal{A} = \left\{ \delta = (\delta_t)_{\{0 \le t \le T\}} \mid \delta \text{ is } \mathcal{F} - \text{predictable and } \mathbb{E} \left[\sup_{0 \le t \le T} (\delta_t)^2 \right] < \infty \right\}.$$

J is not a quasi-convex functional

The value function is not quasiconvex





Figure: Value function for "half-constant" controls. Parameters: T = 1, $\alpha = 0.5$, $\gamma = 0.01$, $\lambda = 100$, $\mu = 1$, and $\sigma = 1$.



Next steps:

- Gâteaux derivative
- Existence and uniqueness
- Global minimum



Let $w, \delta \in A$. The Gâteaux derivative of $J(\delta)$ in the direction of w, is defined as

$$\langle \mathcal{D} J(\delta), w \rangle = \lim_{\epsilon \to 0} \frac{J(\delta + \epsilon w) - J(\delta)}{\epsilon}.$$

Theorem

The functional J is everywhere Gâteaux differentiable and

$$\langle \mathcal{D} J(\delta), w \rangle = \mathbb{E} \left[\int_0^T w_t \phi_t(\delta_t) \left[\delta_t - 2 \gamma \mathbb{E}_{t^-} \left[D_T^{\delta} \right] - (\gamma + \alpha) \right] dA_t \right]$$



The Gâteaux derivative vanishes in all directions if and only if there is $\delta^* \in A$ that satisfies the FBSDE

$$\delta_t^* = 2\gamma \mathbb{E}_{t^-} \left[D_T^{\delta^*} \right] + \gamma + \alpha,$$

almost everywhere in $[0, T] \times \Omega$.



Next steps:

- Existence and Uniqueness of
 - BSDE
 - SDE
 - FBSDE

The FBSDE



The FBSDE

$$\delta_{t} = 2 \gamma \mathbb{E}_{t^{-}} \left[D_{T}^{\delta} \right] + \gamma + \alpha , \qquad (1)$$
$$D_{t}^{\delta} = \int_{0}^{t} \int_{\mathbb{R}} G(\delta_{s} - z) p(\mathrm{d}z, \mathrm{d}s) , \qquad D_{0}^{\delta} = 0 ,$$

with $\delta \in \mathcal{A}$ and $D \in \mathcal{C}$, is equivalent to the FBSDE

$$\tilde{\delta}_{t} = 2 \boldsymbol{\gamma} \mathbb{E}_{t} \left[\int_{t}^{T} \int_{\mathbb{R}} G(\tilde{\delta}_{s^{-}} + 2 \boldsymbol{\gamma} D_{s^{-}}^{\tilde{\delta}} - z) p(\mathrm{d}z, \mathrm{d}s) \right] + \boldsymbol{\gamma} + \boldsymbol{\alpha}, \qquad (2)$$
$$D_{t}^{\tilde{\delta}} = \int_{0}^{t} \int_{\mathbb{R}} G(\tilde{\delta}_{s^{-}} + 2 \boldsymbol{\gamma} D_{s^{-}}^{\tilde{\delta}} - z) p(\mathrm{d}z, \mathrm{d}s), \qquad D_{0}^{\tilde{\delta}} = 0,$$

The BSDE



Theorem Fix $V \in A$. Define Φ_t as

$$\Phi_t(x) = \int_{-\infty}^x \phi_t(y) \mathrm{d}y,$$

and take Φ_t to be uniformly Lipschitz (in $[0, T] \times \Omega$) with constant k, and let $\overline{\lambda}$ be an upper bound of the stochastic intensity λ . The functional $\Psi : \mathcal{A} \to \mathcal{A}$ has a unique fixed point:

$$\Psi(\boldsymbol{U})_t = 2\gamma \mathbb{E}_t \left[\int_t^T \int_{\mathbb{R}} G(\boldsymbol{U}_{s^-} + 2\gamma V_{s^-} - z) p(\mathrm{d}z, \mathrm{d}s) \right] + \gamma + \boldsymbol{\alpha}, \quad \boldsymbol{V} \in \mathcal{A}.$$



Fix $\mathbf{U} \in \mathcal{A}$. Let the distribution function Φ_t be uniformly Lipschitz with constant k and let $\overline{\lambda}$ be an upper bound of the stochastic intensity. The functional $\Theta: \mathcal{C} \to \mathcal{C}$ has a unique fixed point:

$$\Theta(\mathbf{V})_t = \int_0^t \int_{\mathbb{R}} G(\mathbf{U}_{s^-} + 2\gamma \mathbf{V}_{s^-} - z) p(\mathrm{d}z, \mathrm{d}s), \qquad \mathbf{U} \in \mathcal{C}.$$



Let the distribution function Φ_t be uniformly Lipschitz with constant k, such that $k T \overline{\lambda} (\max\{1, 2\gamma\})^2 < 1$. There exits unique solution to the system of FBSDEs, i.e. the functional $\Upsilon : C \otimes C \rightarrow C \otimes C$ defined by

$$\Upsilon(\boldsymbol{U},\boldsymbol{V})_{t} = \begin{pmatrix} H(\boldsymbol{U},\boldsymbol{V})_{t} \\ I(\boldsymbol{U},\boldsymbol{V})_{t} \end{pmatrix} = \begin{pmatrix} 2\gamma \mathbb{E}_{t} \left[\int_{t}^{T} \int_{\mathbb{R}} G(\boldsymbol{U}_{s^{-}} + 2\gamma \boldsymbol{V}_{s^{-}} - z) p(\mathrm{d}z,\mathrm{d}s) \right] + \gamma \\ \int_{0}^{t} \int_{\mathbb{R}} G(\boldsymbol{U}_{s^{-}} + 2\gamma \boldsymbol{V}_{s^{-}} - z) p(\mathrm{d}z,\mathrm{d}s) \end{pmatrix}$$

with

$$\|\Upsilon(\boldsymbol{U},\boldsymbol{V})\|_{\mathcal{C}\otimes\mathcal{C}} = \|H(\boldsymbol{U},\boldsymbol{V})\|_{\mathcal{C}} + \|I(\boldsymbol{U},\boldsymbol{V})\|_{\mathcal{C}},$$

has a unique fixed point.



J is continuous.

Lemma

If J has a global minimum $\hat{\delta} \in A$, then

$$\langle \mathcal{D} J(\hat{\delta}), w \rangle \ge 0 \qquad \forall w \in \mathcal{A}.$$

Theorem

If J has a global minimum $\hat{\delta} \in \mathcal{A}$, then

$$\hat{\delta} = \delta^*$$
 a.e. in $[0, T] \times \Omega$.



Theorem Let $\tilde{\delta}_t = h(t, D_t^{\tilde{\delta}}, \lambda_t)$. The function h (hence $\tilde{\delta}$ and δ^*) satisfies the PIDE

$$0 = \partial_t h(t, D, \lambda) + \mathcal{L}_t^{\lambda} h(t, D, \lambda) + \left(\int_{h(t, D, \lambda)}^{\infty} \lambda \phi_t(z) \, \mathrm{d}z \right) (h(t, D+1, \lambda) - h(t, D, \lambda)) , \qquad (3)$$

with boundary and terminal conditions

$$\lim_{D\to\infty} h(t,D,\lambda) = \infty \quad \text{and} \quad h(T,D,\lambda) = 2\gamma D + \gamma + \alpha.$$

Here, $\mathcal{L}_t^{\lambda}h(t, D, \lambda)$ is the infinitesimal generator of the arrival intensity process λ acting on the function h.



Numerical experiments:

- Take $\lambda = 100$ for all $t \in [0, T]$, so $T_n T_{n-1} \sim \exp(100)$. And $Z_n \sim \mathcal{N}(0.2, 1)$.
- Model parameters are α = 0, γ = 0.07 (when not specified).

Results Optimal control





Figure: Sample paths for the optimal discretion δ^* (top left panel), number of missed trades *D* (lower left panel), cost of strategy *C* (top right panel), and number of trade attempts *N* (lower right panel) for three simulations of the MPP.

Results Optimal control





Figure: Top left panel: Histogram of the extra cost per filled trade $(C_T/(N_T - D_T))$ for various values of γ . Top right panel: Histogram of the cost (C_T) of the strategy for various values of γ . Bottom left panel: Histogram of the number of misses (D_T) for various values of γ . Bottom right panel: Histogram of percentage of misses (D_T/N_T) for various values of γ .

Results Optimal control





Figure: The left-hand side corresponds to fixed discretion strategies and the right-hand corresponds to the strategy δ^* when changing γ . The top panels plot $\mathbb{P}(D_T < 0.1 N_T)$ and the bottom panels the $\mathbb{E}[C_T]$. The marker is placed where the smallest value of $\mathbb{E}[C_T]$ is when $\mathbb{P}(D_T < 0.1 N_T) \ge 0.99$. The model parameters are: $\lambda = 100$, $\alpha = 0$, and $Z_n \sim \mathcal{N}(0.2, 1)$ for every *n*.



Summary:

- We found a global minimum for the tradeoff between costs and misses
- > This framework can be applied to other problems (e.g. "last look")
- We proved existence and uniqueness of a new FBSDE



- Compute the "shadow price of latency" in a similar way to Cartea, Á, and Sánchez-Betancourt, L. (2018)
- Include a latent processes modulating the shocks.



Preliminary results





The framework developed in the previous model can still be applied for HMM modulating the changes in price. Empirical experiments using FX data over April 2019 for EUR/USD show three distinguishable regimes.

Table: Probability distribution of Z_n under each regime (3 regimes in total).

| Y/Z | -5 | -4 | -3 | -2 | -1 | 0 | 1 | 2 | 3 | 4 | 5 |
|-----|------|------|------|-------|-------|-------|-------|------|------|------|------|
| 1 | 3.5% | 3.0% | 5.6% | 10.3% | 15.4% | 26.3% | 14.5% | 9.5% | 5.5% | 3.0% | 3.5% |
| 2 | 0.0% | 0.1% | 0.6% | 3.3% | 12.5% | 66.4% | 12.7% | 3.3% | 0.8% | 0.2% | 0.0% |
| 3 | 0.0% | 0.0% | 0.1% | 0.3% | 1.4% | 96.2% | 1.5% | 0.3% | 0.1% | 0.0% | 0.0% |



Similarly, the transition matrix for the latent process Y is given by

Table: Transition probability matrix for Y.

| Y/Y | 1 | 2 | 3 |
|-----|-------|-------|-------|
| 1 | 77.4% | 22.6% | 0.0% |
| 2 | 5.6% | 86.8% | 7.6% |
| 3 | 3.1% | 3.5% | 93.4% |







Figure: Optimal strategy under each regime. Left panel is regime one, middle panel is for regime two, and right panel is for regime three.



Thank you for your time!

Details can be found in SSRN:

- Cartea, Jaimungal, S-B. (2019) Latency and Liquidity Risk
- Cartea, S-B. (2018) The Shadow Price of Latency: Improving Intraday Fill Ratios in Foreign Exchange Markets