Algorithmic Trading PIMS Summer School 2016 Market Making

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The Limit Order Book

The limit order book is a record of collective interest to buy or sell certain quantities of an asset at a certain price.

Buy Orders		Sell Orders	
Price	Volume	Price	Volume
60.00	80	60.10	75
59.90	100	60.20	75
59.80	90	60.30	50

Graphical representation of the limit order book:



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Market Orders

• An incoming market order lifts limit orders from the book.



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Agent's Goal

Optimally place limit orders in the limit order book (LOB)



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Market Making

- The market maker's problem is to find prices at which to post limit buy/sell orders to profit from round-trip trades
- ► The **benchmark models**: Ho & Stoll (81), Avellanda & Stoikov (08) Cartea & Jaimungal (12), and others
- Need to account for
 - market order arrival rate
 - Probability that market maker is filled at a given level
 - midprice dynamics

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Market Making

• Midprice $dS_t = \sigma \, dW_t$, $\sigma > 0$,

• δ^{\pm} depth at which the agent posts LOs:

- Sell LOs are posted at a price of $S_t + \delta_t^+$
- Buy LOs at $S_t \delta_t^-$
- M[±] Poisson arrival of other participants' buy (+) and sell (−) MOs which arrive at with intensities λ[±],
- ▶ $N^{\delta,\pm}$ counting processes for the agent's filled sell (+) and buy (-) LOs,
- ► Conditional on an MO arrival, the LO is filled with probability $e^{-\kappa^{\pm} \delta_t^{\pm}}$ with $\kappa^{\pm} \ge 0$

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Market Maker's Control Problem

The MM's performance criteria is

$$H^{\delta}(t,x,S,q) = \mathbb{E}_{t,x,q,S}\left[X_T^{\delta} + Q_T^{\delta}(S_T - \boldsymbol{\alpha} Q_T^{\delta}) - \boldsymbol{\phi} \int_t^T (Q_u^{\delta})^2 du\right],$$

 $\pmb{lpha} \geq$ 0, and $\pmb{\phi} \geq$ 0. Value function is

$$H(t,x,S,q) = \sup_{\delta^{\pm}\in\mathcal{A}} H^{\delta}(t,x,S,q),$$

and inventory capped: above by $\overline{q} > 0$ and below by q < 0.

X^δ cash process

$$dX_t^{\delta} = \left(S_t + \delta_t^+\right) \, d\mathbf{N}_t^{\delta,+} - \left(S_t - \delta_t^-\right) \, d\mathbf{N}_t^{\delta,-}$$

• Q^{δ} inventory process and satisfies

$$Q_t^{\delta} = \mathbf{N}_t^{\delta,-} - \mathbf{N}_t^{\delta,+}$$

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DPE

A DPP holds and the value function satisfies the DPE

$$\begin{split} \mathbf{0} &= \partial_t H + \underbrace{\frac{1}{2} \sigma^2 \partial_{SS} H}_{\text{midprice diffusion}} - \underbrace{\phi q^2}_{\text{running inv penalty}} \\ &+ \lambda^+ \sup_{\delta^+} \left\{ \underbrace{\mathbf{e}^{-\kappa^+ \delta^+}}_{\text{Prob Filled sell LO}} \left(H(t, \mathbf{x} + (\mathbf{S} + \delta^+), \mathbf{q} - \mathbf{1}, S) - H \right) \right\} \mathbbm{1}_{q > \underline{q}} \\ &+ \lambda^- \sup_{\delta^-} \left\{ \underbrace{\mathbf{e}^{-\kappa^- \delta^-}}_{\text{Prob Filled buy LO}} \left(H\left(t, \mathbf{x} - (\mathbf{S} - \delta^-), \mathbf{q} + \mathbf{1}, S\right) - H \right) \right\} \mathbbm{1}_{q < \overline{q}}, \end{split}$$

where 1 is the indicator function, and with terminal condition

$$H(T, x, S, q) = x + q (S - \alpha q).$$

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Solving HJB

▶ Make an ansatz for *H*. In particular, write

$$H(t, x, q, S) = x + q S + h(t, q).$$

- first term is accumulated cash
- second term is the book value of the inventory marked-to-market
- Iast term is the added value from following an optimal market making strategy up to T.

$$\begin{split} \phi q^2 &= \partial_t h(t,q) + \lambda^+ \sup_{\delta^+} \left\{ \mathbf{e}^{-\kappa^+ \delta^+} \left(\delta^+ + h(t,q-1) - h(t,q) \right) \right\} \, \mathbb{1}_{q > \underline{q}} \\ &+ \lambda^- \sup_{\delta^-} \left\{ \mathbf{e}^{-\kappa^- \delta^-} \left(\delta^- + h(t,q+1) - h(t,q) \right) \right\} \, \mathbb{1}_{q < \overline{q}} \,, \end{split}$$

with terminal condition $h(T,q) = -\alpha q^2$.

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Optimal Controls

Then the optimal depths in feedback form are given by

$$\delta^{+,*}(t,q)=rac{1}{\kappa^+}-h(t,q-1)+h(t,q)\,,\quad q
eq q\,,$$
 (1a)

$$\delta^{-,*}(t,q) = rac{1}{\kappa^-} - h(t,q+1) + h(t,q), \quad q \neq \overline{q},$$
 (1b)

and boundaries $\delta^{+,*}(t,\overline{q}) = +\infty$ and $\delta^{-,*}(t,\underline{q}) = +\infty$.

Substituting the optimal controls into the DPE we obtain

$$\phi q^{2} = \partial_{t} h(t,q) + \frac{\lambda^{+}}{\kappa^{+}} e^{-1} e^{-\kappa^{+}(-h(t,q-1)+h(t,q))} \mathbb{1}_{q \geq \underline{q}} + \frac{\lambda^{-}}{\kappa^{-}} e^{-1} e^{-\kappa^{-}(-h(t,q+1)+h(t,q))} \mathbb{1}_{q < \overline{q}}.$$
(2)

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Symmetric fill probability

Analytical solution if $\kappa = \kappa^+ = \kappa^-$:

$$h(t,q) = rac{1}{\kappa} \log \omega(t,q),$$

and stack $\omega(t,q)$ into a vector

$$oldsymbol{\omega}(t,q) = ig[\omega(t,\overline{q}),\,\omega(t,\overline{q}-1),\dots,\,\omega(t,\underline{q})ig]'$$
 .

Now, let **A** denote the $(\overline{q} - \underline{q} + 1)$ -square matrix whose rows are labeled from \overline{q} to q and whose entries are given by

$$\mathbf{A}_{i,q} = \begin{cases} -\phi \kappa \, q^2 \,, & i = q \,, \\ \lambda^+ \, e^{-1} \,, & i = q - 1 \,, \\ \lambda^- \, e^{-1} \,, & i = q + 1 \,, \\ 0 \,, & \text{otherwise,} \end{cases}$$
(3)

with terminal and boundary conditions $\omega(T,q) = e^{-\alpha \kappa q^2}$.

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Then,

$$\omega(t) = e^{\mathbf{A}(\tau - t)} \mathbf{z} \,, \tag{4}$$

where \mathbf{z} is a $(\overline{q} - \underline{q} + 1)$ -dim vector where each component is $z_j = e^{-\alpha \kappa j^2}$, $j = \overline{q}, \dots, \underline{q}$. Inserting the controls (1) into the DPE

equation (2) and writing $h(t,q) = \frac{1}{\kappa} \log \omega(t,q)$, after some straightforward computations, one finds that $\omega(t,q)$ satisfy the coupled system of equations

$$\partial_t \boldsymbol{\omega}(t) + \mathbf{A} \boldsymbol{\omega}(t) = \mathbf{0}.$$
 (5)

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Optimal Postings

Optimal postings $\phi = 0.001$



Figure : The optimal depths as a function of time for various inventory levels, T = 30, $\lambda^{\pm} = 1$, $\kappa^{\pm} = 100$, $\overline{q} = -\underline{q} = 3$, $\alpha = 0.0001$, $\sigma = 0.01$, $S_0 = 100$.

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Optimal postings $\phi = 0.02$



Figure : The optimal depths as a function of time for various inventory levels, T = 30, $\lambda^{\pm} = 1$, $\kappa^{\pm} = 100$, $\overline{q} = -\underline{q} = 3$, $\alpha = 0.0001$, $\sigma = 0.01$, $S_0 = 100$.

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Mean reversion in inventory

Given the pair of optimal strategies $\delta^+(t, q), \delta^-(t, q)$, the expected drift in inventories q_t is given by

$$\mu(t,q) \triangleq \lim_{s \downarrow t} \frac{1}{s-t} \mathbb{E} \left[Q_s - Q_t \, | \, Q_{t^-} = q \right] ,$$

= $\lambda^- e^{-\kappa^- \delta^-, *(t,q)} - \lambda^+ e^{-\kappa^+ \delta^+, *(t,q)} .$ (6)

- The drift $\mu(t, q)$ depends on time.
- For the same level of inventory the speed depends on how near of far is the strategy from T

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Inventory



Figure : Long-term inventory level. Model parameters are: $\lambda^{\pm} = 1$, $\kappa^{\pm} = 100$, $\bar{q} = -\underline{q} = 10$, $\alpha = 0.0001$, $\sigma = 0.01$, $S_0 = 100$, and $\phi = \{2 \times 10^{-3}, 10^{-3}, 5 \times 10^{-4}\}$.

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Figure : Inventory and midprice path. Model parameters are: $\lambda^{\pm} = 1$, $\kappa^{\pm} = 100$, $\overline{q} = -\underline{q} = 10$, $\phi = 0.02$, $\alpha = 0.0001$, $\sigma = 0.01$, $S_0 = 100$.

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Profit and Loss



Figure : P&L and Life Inventory of the optimal strategy for 10,000 simulations, $\lambda^{\pm} = 1$, $\kappa^{\pm} = 100$, $\overline{q} = -\underline{q} = 10$, $\alpha = 0.0001$, $\sigma = 0.01$, and $S_0 = 100$.

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Market Making with No Terminal Penalty

Solving HJB with $\alpha = 0$

Assume no penalties for liquidating inventories at time T. Thus the ansatz is

$$H(t, x, q, S) = x + q S + \mathbf{g(t)}.$$
(7)

Thus,

$$0 = g_t(t) + \lambda^+ \sup_{\delta^+} \left\{ \mathbf{e}^{-\kappa^+ \delta^+} \, \delta^+ \right\} + \lambda^- \sup_{\delta^-} \left\{ \mathbf{e}^{-\kappa^- \delta^-} \, \delta^- \right\} \,,$$

and the optimal postings are:

$$\delta^{*,+} = \frac{1}{\kappa^+}$$

and

$$\delta^{*,-} = rac{1}{\kappa^-} \, .$$

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Solving HJB with $\alpha = 0$

Alternatively note that

- A risk-neutral MM, who does not penalise inventories, seeks to maximise the probability of being filled at every instant in time.
- ► Thus, the MM chooses δ[±] to maximise the expected depth conditional on a market order hitting or lifting the appropriate side of the book: maximises δ[±]e^{-κ[±]δ[±]}. The FOC

$$e^{-\kappa^{\pm}\delta^{\pm}} - \kappa^{\pm}\delta^{\pm}e^{-\kappa^{\pm}\delta^{\pm}} = 0.$$
(8)

Thus, we see that the optimal half spreads are as above.

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