

Algorithmic Trading: Market Making

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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**

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^\pm Poisson arrival of other participants' buy (+) and sell (−) MOs which arrive at with intensities λ^\pm ,
- ▶ $N^{\delta,\pm}$ 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 $\kappa^\pm \geq 0$

Market Maker's Control Problem

The MM's performance criteria is

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

$\alpha \geq 0$, and $\phi \geq 0$. 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 $\bar{q} > 0$ and below by $\underline{q} < 0$.

- ▶ X^δ cash process

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

- ▶ Q^δ inventory process and satisfies

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

A DPP holds and the value function satisfies the DPE

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

where $\mathbb{1}$ is the indicator function, and with terminal condition

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

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
- last term is the added value from following an optimal market making strategy up to T .

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

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

Optimal Controls

Then the optimal depths in feedback form are given by

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

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

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

Substituting the optimal controls into the DPE we obtain

$$\begin{aligned} \phi q^2 = \partial_t h(t, q) &+ \frac{\lambda^+}{\kappa^+} e^{-1} e^{-\kappa^+(-h(t, q-1) + h(t, q))} \mathbb{1}_{q > \underline{q}} \\ &+ \frac{\lambda^-}{\kappa^-} e^{-1} e^{-\kappa^-(-h(t, q+1) + h(t, q))} \mathbb{1}_{q < \bar{q}}. \end{aligned} \quad (2)$$

Symmetric fill probability

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

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

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

$$\boldsymbol{\omega}(t, q) = [\omega(t, \bar{q}), \omega(t, \bar{q} - 1), \dots, \omega(t, \underline{q})]'.$$

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

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

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

Then,

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

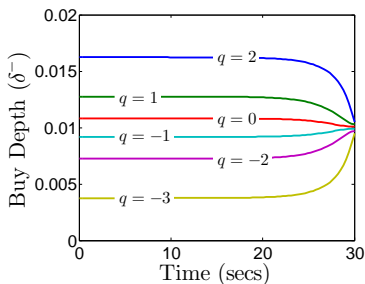
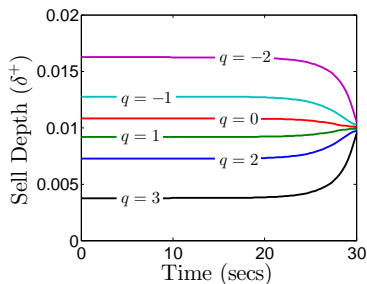
where \mathbf{z} is a $(\bar{q} - \underline{q} + 1)$ -dim vector where each component is $z_j = e^{-\alpha_\kappa j^2}$, $j = \bar{q}, \dots, \underline{q}$. Inserting the controls (1) into the DPE

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

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

Optimal Postings

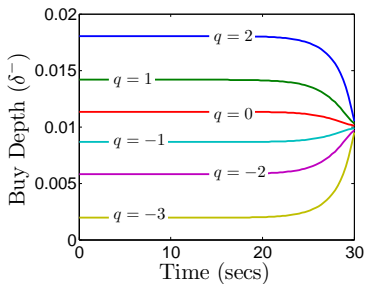
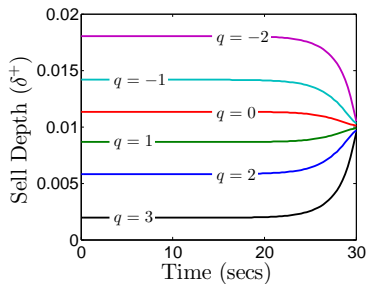
Optimal postings $\phi = 0.001$



(a) $\phi = 0.001$

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

Optimal postings $\phi = 0.02$



(a) $\phi = 0.02$

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

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

$$\begin{aligned}\mu(t, q) &\triangleq \lim_{s \downarrow t} \frac{1}{s - t} \mathbb{E}[Q_s - Q_t \mid Q_{t-} = q], \\ &= \lambda^- e^{-\kappa^- \delta^{-,*}(t, q)} - \lambda^+ e^{-\kappa^+ \delta^{+,*}(t, q)}.\end{aligned}\tag{6}$$

- ▶ 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

Inventory

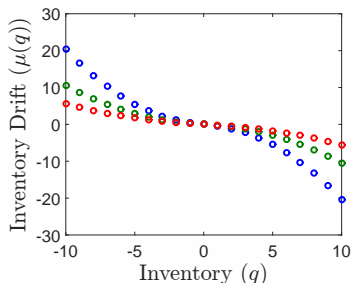
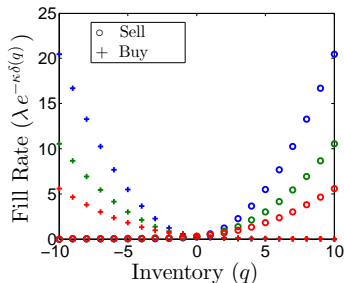


Figure: Long-term inventory level. Model parameters are: $\lambda^{\pm} = 1$, $\kappa^{\pm} = 100$, $\bar{q} = -q = 10$, $\alpha = 0.0001$, $\sigma = 0.01$, $S_0 = 100$, and $\phi = \{2 \times 10^{-3}, 10^{-3}, 5 \times 10^{-4}\}$.

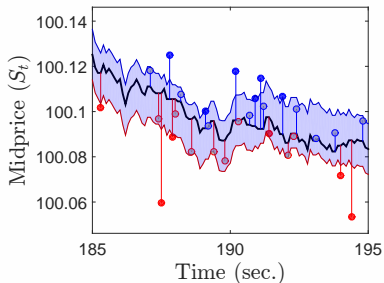
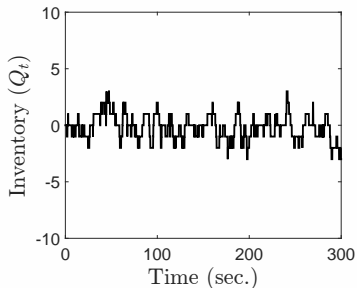


Figure: Inventory and midprice path. Model parameters are: $\lambda^\pm = 1$, $\kappa^\pm = 100$, $\bar{q} = -\underline{q} = 10$, $\phi = 0.02$, $\alpha = 0.0001$, $\sigma = 0.01$, $S_0 = 100$.

Profit and Loss

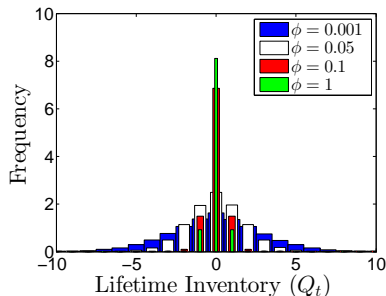
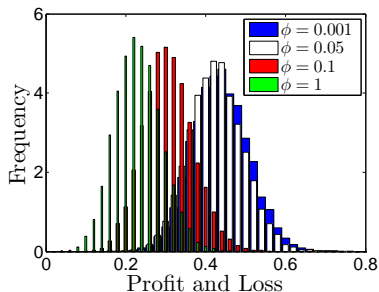


Figure: P&L and Life Inventory of the optimal strategy for 10,000 simulations, $\lambda^\pm = 1$, $\kappa^\pm = 100$, $\bar{q} = -\underline{q} = 10$, $\alpha = 0.0001$, $\sigma = 0.01$, and $S_0 = 100$.

Market Making with No Terminal Penalty

Solving HJB with $\alpha = 0$

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

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

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

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

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^\pm to maximise the expected depth conditional on a market order hitting or lifting the appropriate side of the book: maximises $\delta^\pm e^{-\kappa^\pm \delta^\pm}$. The FOC

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

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