

High-Frequency Algorithmic Trading with Momentum and Order Imbalance

My goal is to establish and solve the stochastic optimal control problem that captures the momentum and order imbalance dynamics of the Limit Order Book (LOB). The solution will yield an optimal trading strategy that will permit statistical arbitrage of the underlying stock, which will then be backtested on historical data.

Progress Timeline

DATE	THESIS	STA4505
Dec-2014	• Complete CTMC calibration	
Dec-2014	• Backtest naive strategies based on CTMC	
Jan-May	• Study stochastic controls: ECE1639, STA4505	
Jun-5	• Establish models	Exam Study
Jun-12	• Establish performance criteria	Exam Study
Jun-15	• Derive DPP/DPE	EXAM
Jun-19	• Derive DPP/DPE	
Jun-26	• Derive continuous-time equations	
Jul-3	• Derive discrete-time equations	
Jul-10	• Set up MATLAB numerical integration	
Jul-17	• Integrate functions and plot dynamics	Integrate and analyze
Jul-24	• More dynamics, and calib/choose parameters	
Jul-31	• Backtest on historical data	Simulate results
Aug-7	• More backtesting, comparing with previous	
Aug-14	• Dissertation writeup / buffer	Project writeup
Aug-21	• Dissertation writeup / buffer	
Aug-28	• Dissertation writeup	Presentation

Whiteboard Inspirational Quote of the Week

I'm sorry to say that the subject I most disliked was mathematics. I have thought about it. I think the reason was that mathematics leaves no room for argument. If you made a mistake, that was all there was to it.

- Malcolm X

The Academic Week in Review

Reminder of our processes (a little bit of abuse of notation going on):

$\mathbf{z}_k = (\rho_k, \Delta_k)$ - 2-D time-homogenous Markov Chain with transition probabilities P_{ij} , where $\rho_k \in \Gamma$ and Γ represents the set of imbalance bins, and $\Delta_k = \text{sgn}(s_k - s_{k-1}) \in \{-1, 0, 1\}$.

State $\vec{x}_k =$	$\begin{pmatrix} x_k \\ s_k \\ \mathbf{z}_k \\ q_k \end{pmatrix}$	cash stock price Markov chain state, as above inventory
Control $\vec{u}_k =$	$\begin{pmatrix} \delta_k^+ \\ \delta_k^- \\ M_k^+ \\ M_k^- \end{pmatrix}$	bid posting depth ask posting depth buy MO - binary control sell MO - binary control
Random $\vec{w}_k =$	$\begin{pmatrix} K_k^+ \\ K_k^- \\ \omega_k \end{pmatrix}$	other agent buy MOs - binary other agent sell MOs - binary random variable uniformly distributed on $[0,1]$

We'll write the evolution of the Markov chain as a function of the current state and a uniformly distributed random variable ω :¹

$$\mathbf{z}_{k+1} = T(\mathbf{z}_k, \omega_k) = \sum_{i=0}^{|\Gamma|} i \cdot \mathbb{1}_{(\sum_{j=0}^{i-1} P_{\mathbf{z}_k, j}, \sum_{j=0}^i P_{\mathbf{z}_k, j}]}(\omega_k) \quad (1)$$

Here $\mathbb{1}_A(\omega) = \begin{cases} 1 & \text{if } \omega \in A \\ 0 & \text{if } \omega \notin A \end{cases}$, and hence \mathbf{z}_{k+1} is assigned to the value i for which ω_k is in the indicated interval of probabilities.

Our Markovian state evolution function f , given by $\vec{x}_{k+1} = f(\vec{x}_k, \vec{u}_k, \vec{w}_k)$, can be written explicitly as

$$\begin{pmatrix} x_{k+1} \\ s_{k+1} \\ \mathbf{z}_{k+1} \\ q_{k+1} \end{pmatrix} = \begin{pmatrix} x_k + (s_k + \pi + \delta_k^-)L_k^- - (s_k - \pi - \delta_k^+)L_k^+ + (s_k - \pi)M_k^- - (s_k + \pi)M_k^+ \\ s_k + \eta_{k, \mathbf{z}_k} T(\mathbf{z}_k, \omega_k)^{(2)} \\ T(\mathbf{z}_k, \omega_k) \\ q_k + L_k^+ - L_k^- + M_k^+ - M_k^- \end{pmatrix} \quad (2)$$

¹Borrowed from ECE1639 notes.

Clearly, the cash process at a subsequent is equal to the cash at the previous step, plus the costs of profits of executing market or limit orders. There are two noteworthy observations regarding this formulation of the evolution function. First, note that the price paid/received for limit orders depends on the stock price at time k . This implies that at k , if the agent posts a sell limit order, and the binary random variable L_k^- (which depends on the binary random variable M_k^+) is equal to 1, then the agent's order is filled "between timesteps" k and $k+1$, but using the price at time k . Second, since the second dimension $T(\mathbf{z}_k, \omega_k)^{(2)} = \Delta_{k+1} = \text{sgn}(s_{k+1} - s_k)$ determines the directionality of the price jump between times k and $k+1$, multiplying it by the random variable η_{k, \mathbf{z}_k} determines the size of the price change.

Max Terminal Wealth (Discrete)

Following traditional dynamic programming, we introduce the value function V_k^u . Here, our objective is to maximize the terminal wealth performance criteria given by

$$V_k^u(x, s, \mathbf{z}, q) = \mathbb{E}[W_T^u] = \mathbb{E}_{k, x, s, \mathbf{z}, q} [X_T^u + Q_T^u(S_T - \text{sgn}(Q_T^u)\pi) - \alpha(Q_T^u)^2] \quad (3)$$

where, as before, the notation $\mathbb{E}_{k, x, s, \mathbf{z}, q}[\cdot]$ represents the conditional expectation

$$\mathbb{E}[\cdot \mid X_k = x, S_k = s, \mathbf{Z}_k = \mathbf{z}, Q_k = q]$$

In this case, our dynamic programming equations (DPEs) are given by

$$V_k(x, s, \mathbf{z}, q) = \sup_{\mathbf{u}} \{ \mathbb{E}_{\mathbf{w}} [V_{k+1}(f((x, s, \mathbf{z}, q), \mathbf{u}, \mathbf{w}_k))] \} \quad (4)$$

$$V_T(x, s, \mathbf{z}, q) = \sup_{\mathbf{u}} \{ \mathbb{E} [x + q(s - \text{sgn}(q)\pi) - \alpha q^2] \} \quad (5)$$

where expectation is with respect to the random vector \mathbf{w}_k .

To simplify the DPEs, we introduce a now familiar ansatz:

$$V_k = x + q(s - \text{sgn}(q)\pi) + h_k(\mathbf{z}, q) \quad (6)$$

with boundary condition $h_k(\mathbf{z}, 0) = 0$ and terminal condition $h_T(\mathbf{z}, q) = -\alpha q^2$. Substituting this into the DPE, we obtain

$$\begin{aligned} 0 &= \sup_{\mathbf{u}} \{ \mathbb{E}_{\mathbf{w}} [V_{k+1}(f((x, s, \mathbf{z}, q), \mathbf{u}, \mathbf{w}_k))] \} - V_k(x, s, \mathbf{z}, q) \\ &= \sup_{\mathbf{u}} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ + (s - \pi) M_k^- - (s + \pi) M_k^+ \right. \right. \\ &\quad + (L_k^+ - L_k^- + M_k^+ - M_k^-) \\ &\quad \times (s + \eta_{0, \mathbf{z}} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^- + M_k^+ - M_k^-) \pi) \\ &\quad + q (\eta_{0, \mathbf{z}} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^- + M_k^+ - M_k^-) - \text{sgn}(q)) \pi) \\ &\quad \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^- + M_k^+ - M_k^-) - h_k(\mathbf{z}, q) \right] \right\} \end{aligned} \quad (7)$$

Since our buy/sell market order controls are binary, the supremum over the control vector \mathbf{u} can be treated as a simultaneous supremum over δ^\pm and maximum over the four possible values for M^\pm . Notably, however, a quick substitution shows that the case where $M^+ = M^- = 1$ is not possible as it is always strictly 2π less in value than the case of only limit orders, where $M^+ = M^- = 0$. This should be evident, as buying and selling with market orders in a single timestep yields a guaranteed loss as the agent is forced to cross the spread. Thus, our DPE takes the form:

$$\begin{aligned}
0 = \max \Bigg\{ & \sup_{\delta^\pm} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ \right. \right. \\
& + (L_k^+ - L_k^-) (s + \eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^-) \pi) \\
& + q (\eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^-) - \text{sgn}(q)) \pi) \\
& \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^-) - h_k(\mathbf{z}, q) \right] \right\} ; \\
& \sup_{\delta^\pm} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ - (s + \pi) \right. \right. \\
& + (L_k^+ - L_k^- + 1) (s + \eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^- + 1) \pi) \\
& + q (\eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^- + 1) - \text{sgn}(q)) \pi) \\
& \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^- + 1) - h_k(\mathbf{z}, q) \right] \right\} ; \\
& \sup_{\delta^\pm} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ + (s - \pi) \right. \right. \\
& + (L_k^+ - L_k^- - 1) (s + \eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^- - 1) \pi) \\
& + q (\eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^- - 1) - \text{sgn}(q)) \pi) \\
& \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^- - 1) - h_k(\mathbf{z}, q) \right] \right\} \Bigg\} \tag{8}
\end{aligned}$$

We'll begin by concentrating on the first case where $M^+ = M^- = 0$. Thus, we want to solve

$$\begin{aligned}
& \sup_{\delta^\pm} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ \right. \right. \\
& + (L_k^+ - L_k^-) (s + \eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^-) \pi) \\
& + q (\eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^-) - \text{sgn}(q)) \pi) \\
& \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^-) - h_k(\mathbf{z}, q) \right] \right\} \tag{9}
\end{aligned}$$

You might recall that things got pretty messy pretty fast. Previously we had set up the problem such that at each timestep k there can be multiple other agents' market orders (K_k^+) arriving, and these were Poisson distributed. For each arriving order, the probability of our posted limit order being filled was $e^{-\kappa \delta^-}$. We're going to modify this slightly. Keeping the market orders as

Poisson distributed, we have that $\mathbb{P}[K_k^+ = 0] = \frac{e^{-\lambda(\mathbf{z})\Delta t}(\lambda(\mathbf{z})\Delta t)^0}{0!} = e^{-\lambda(\mathbf{z})\Delta t}$, and so the probability of seeing some positive number of market orders is

$$\mathbb{P}[K_k^+ > 0] = 1 - e^{-\lambda(\mathbf{z})\Delta t} \quad (10)$$

Now we make the simplified assumption that the *aggregate* of the orders walks the limit order book to a depth of p_k , and if $p_k > \delta^-$, then our sell limit order is lifted. Thus we have the following preliminary results:

$$\begin{aligned} \mathbb{P}[L_k^- = 1 | K_k^+ > 0] &= e^{-\kappa\delta^-} \\ \mathbb{P}[L_k^- = 0 | K_k^+ > 0] &= 1 - e^{-\kappa\delta^-} \\ \mathbb{E}[L_k^-] &= \mathbb{P}[L_k^- = 1 | K_k^+ > 0] \cdot \mathbb{P}[K_k^+ > 0] \\ &= (1 - e^{-\lambda(\mathbf{z})\Delta t})e^{-\kappa\delta^-} \end{aligned}$$

For ease of notation, we'll write the probability of the $L_k^- = 1$ event as $p(\delta^-)$. This gives us the additional results:

$$\begin{aligned} \mathbb{P}[L_k^- = 1] &= p(\delta^-) = \mathbb{E}[L_k^-] \\ \mathbb{P}[L_k^- = 0] &= 1 - p(\delta^-) \\ \partial_{\delta^-} \mathbb{P}[L_k^- = 1] &= -\kappa p(\delta^-) \\ \partial_{\delta^-} \mathbb{P}[L_k^- = 0] &= \kappa p(\delta^-) \end{aligned}$$

Contributing to last week's frustration, a la our man Malcolm X, was an error in calculation. I had mistakenly calculated $\mathbb{E}[L_k^- \text{sgn}(q + L_k^+ - L_k^-)] = \mathbb{E}[L_k^-] \mathbb{E}[\text{sgn}(q + L_k^+ - L_k^-)]$, as if to say they are independent, but of course they are not. This is fixed below. On the contrary, $\mathbb{E}[L_k^- \eta_{0,\mathbf{z}} T(\mathbf{z}, \omega)^{(2)}]$ was correctly treated as the product of independent random variables. We have that $\mathbb{E}[L_k^-] = (1 - e^{-\lambda(\mathbf{z})\Delta t})e^{-\kappa\delta^-}$ is clearly dependent on \mathbf{z} , but these expectations are over the vector of random variables $\mathbf{w} = (K^+, K^-, \omega)$ and are evaluated at a given point \mathbf{z} .

Let's pre-compute some of the terms that we'll encounter in the supremum, namely the expectations of the random variables.

$$\begin{aligned} \mathbb{E}[\text{sgn}(q + L_k^+ - L_k^-)] &= \mathbb{P}[L_k^- = 1] \cdot \mathbb{P}[L_k^+ = 1] \cdot \text{sgn}(q) \\ &\quad + \mathbb{P}[L_k^- = 1] \cdot \mathbb{P}[L_k^+ = 0] \cdot \text{sgn}(q - 1) \\ &\quad + \mathbb{P}[L_k^- = 0] \cdot \mathbb{P}[L_k^+ = 1] \cdot \text{sgn}(q + 1) \\ &\quad + \mathbb{P}[L_k^- = 0] \cdot \mathbb{P}[L_k^+ = 0] \cdot \text{sgn}(q) \\ &= p(\delta^-)p(\delta^+) \text{sgn}(q) \\ &\quad + p(\delta^-)(1 - p(\delta^+)) \text{sgn}(q - 1) \\ &\quad + (1 - p(\delta^-))p(\delta^+) \text{sgn}(q + 1) \\ &\quad + (1 - p(\delta^-))(1 - p(\delta^+)) \text{sgn}(q) \\ &= \text{sgn}(q) [1 - p(\delta^+) - p(\delta^-) + 2p(\delta^+)p(\delta^-)] \end{aligned}$$

$$\begin{aligned}
& + \operatorname{sgn}(q-1)[p(\delta^-) - p(\delta^+)p(\delta^-)] \\
& + \operatorname{sgn}(q+1)[p(\delta^+) - p(\delta^+)p(\delta^-)] \\
& = \begin{cases} 1 & q \geq 2 \\ 1 - p(\delta^-)(1 - p(\delta^+)) & q = 1 \\ p(\delta^+) - p(\delta^-) & q = 0 \\ -[1 - p(\delta^+)(1 - p(\delta^-))] & q = -1 \\ -1 & q \leq -2 \end{cases} \quad (11)
\end{aligned}$$

$$= \Phi(q, \delta^+, \delta^-) \quad (12)$$

Similarly:

$$\begin{aligned}
\mathbb{E}[L_k^+ \operatorname{sgn}(q + L_k^+ - L_k^-)] &= \mathbb{P}[L_k^- = 1] \cdot \mathbb{P}[L_k^+ = 1] \cdot \operatorname{sgn}(q) \\
&+ \mathbb{P}[L_k^- = 1] \cdot \mathbb{P}[L_k^+ = 0] \cdot 0 \operatorname{sgn}(q-1) \\
&+ \mathbb{P}[L_k^- = 0] \cdot \mathbb{P}[L_k^+ = 1] \cdot \operatorname{sgn}(q+1) \\
&+ \mathbb{P}[L_k^- = 0] \cdot \mathbb{P}[L_k^+ = 0] \cdot 0 \operatorname{sgn}(q) \\
&= p(\delta^+) [p(\delta^-) \operatorname{sgn}(q) + (1 - p(\delta^-) \operatorname{sgn}(q+1))] \\
&= p(\delta^+) \begin{cases} 1 & q \geq 2 \\ 1 & q = 1 \\ (1 - p(\delta^-)) & q = 0 \\ -p(\delta^-) & q = -1 \\ -1 & q \leq -2 \end{cases} \quad (13)
\end{aligned}$$

$$= p(\delta^+) \Psi(q, \delta^-) \quad (14)$$

and

$$\begin{aligned}
\mathbb{E}[L_k^- \operatorname{sgn}(q + L_k^+ - L_k^-)] &= p(\delta^-) [p(\delta^+) \operatorname{sgn}(q) + (1 - p(\delta^+)) \operatorname{sgn}(q-1)] \\
&= p(\delta^-) \begin{cases} 1 & q \geq 2 \\ p(\delta^+) & q = 1 \\ -(1 - p(\delta^+)) & q = 0 \\ -1 & q = -1 \\ -1 & q \leq -2 \end{cases} \quad (15)
\end{aligned}$$

$$= p(\delta^-) \Upsilon(q, \delta^+) \quad (16)$$

We'll also require the partial derivatives of these expectations, which we can easily compute. Below we'll use the simplified notation Φ_+ to denote the function closely associated with the partial derivative of Φ with respect to δ^+ .

$$\partial_{\delta^-} \mathbb{E}[\operatorname{sgn}(q + L_k^+ - L_k^-)] = \partial_{\delta^-} \Phi(q, \delta^+, \delta^-) = \kappa p(\delta^-) \begin{cases} 0 & q \geq 2 \\ (1 - p(\delta^+)) & q = 1 \\ 1 & q = 0 \\ p(\delta^+) & q = -1 \\ 0 & q \leq -2 \end{cases} \quad (17)$$

$$= \kappa p(\delta^-) \Phi_-(q, \delta^+) \quad (18)$$

$$\partial_{\delta^+} \mathbb{E}[\text{sgn}(q + L_k^+ - L_k^-)] = \partial_{\delta^+} \Phi(q, \delta^+, \delta^-) = \kappa p(\delta^+) \begin{cases} 0 & q \geq 2 \\ -p(\delta^-) & q = 1 \\ -1 & q = 0 \\ -(1 - p(\delta^-)) & q = -1 \\ 0 & q \leq -2 \end{cases} \quad (19)$$

$$= \kappa p(\delta^+) \Phi_+(q, \delta^-) \quad (20)$$

$$\partial_{\delta^-} \mathbb{E}[L_k^+ \text{sgn}(q + L_k^+ - L_k^-)] = \partial_{\delta^-} p(\delta^+) \Psi(q, \delta^-) = \kappa p(\delta^+) p(\delta^-) \begin{cases} 0 & q \geq 2 \\ 0 & q = 1 \\ 1 & q = 0 \\ 1 & q = -1 \\ 0 & q \leq -2 \end{cases} \quad (21)$$

$$= \kappa p(\delta^+) p(\delta^-) \Psi_-(q) \quad (22)$$

$$\partial_{\delta^+} \mathbb{E}[L_k^+ \text{sgn}(q + L_k^+ - L_k^-)] = \partial_{\delta^+} p(\delta^+) \Psi(q, \delta^-) = -\kappa p(\delta^+) \Psi(q, \delta^-) \quad (23)$$

$$\partial_{\delta^-} \mathbb{E}[L_k^- \text{sgn}(q + L_k^+ - L_k^-)] = \partial_{\delta^-} p(\delta^-) \Upsilon(q, \delta^+) = -\kappa p(\delta^-) \Upsilon(q, \delta^+) \quad (24)$$

$$\partial_{\delta^+} \mathbb{E}[L_k^- \text{sgn}(q + L_k^+ - L_k^-)] = \partial_{\delta^+} p(\delta^-) \Upsilon(q, \delta^+) = \kappa p(\delta^+) p(\delta^-) \begin{cases} 0 & q \geq 2 \\ -1 & q = 1 \\ -1 & q = 0 \\ 0 & q = -1 \\ 0 & q \leq -2 \end{cases} \quad (25)$$

$$= \kappa p(\delta^+) p(\delta^-) \Upsilon_+(q) \quad (26)$$

Recalling that we have \mathbf{P} the transition matrix for the Markov Chain \mathbf{Z} , with $\mathbf{P}_{\mathbf{z}, \mathbf{j}} = \mathbb{P}[\mathbf{Z}_{k+1} = \mathbf{j} | \mathbf{Z}_k = \mathbf{z}]$, then we can also write:

$$\begin{aligned} \mathbb{E}[h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^-)] &= \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) [1 - p(\delta^+) - p(\delta^-) + 2p(\delta^+)p(\delta^-)] \right. \\ &\quad + h_{k+1}(\mathbf{j}, q - 1) [p(\delta^-) - p(\delta^+)p(\delta^-)] \\ &\quad \left. + h_{k+1}(\mathbf{j}, q + 1) [p(\delta^+) - p(\delta^+)p(\delta^-)] \right] \end{aligned} \quad (27)$$

and its partial derivatives as

$$\begin{aligned} \partial_{\delta^-} \mathbb{E}[h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^-)] &= \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) [\kappa p(\delta^-) - 2\kappa p(\delta^+)p(\delta^-)] \right. \\ &\quad + h_{k+1}(\mathbf{j}, q - 1) [-\kappa p(\delta^-) + \kappa p(\delta^+)p(\delta^-)] \\ &\quad \left. + h_{k+1}(\mathbf{j}, q + 1) [\kappa p(\delta^+)p(\delta^-)] \right] \end{aligned} \quad (28)$$

$$\begin{aligned}
&= \kappa p(\delta^-) \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) [1 - 2p(\delta^+)] \right. \\
&\quad \left. + h_{k+1}(\mathbf{j}, q-1) [-1 + p(\delta^+)] \right. \\
&\quad \left. + h_{k+1}(\mathbf{j}, q+1) [p(\delta^+)] \right] \tag{29}
\end{aligned}$$

$$\begin{aligned}
\partial_{\delta^+} \mathbb{E}[h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^-)] &= \kappa p(\delta^+) \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) [1 - 2p(\delta^-)] \right. \\
&\quad \left. + h_{k+1}(\mathbf{j}, q-1) [p(\delta^-)] \right. \\
&\quad \left. + h_{k+1}(\mathbf{j}, q+1) [-1 + p(\delta^-)] \right] \tag{30}
\end{aligned}$$

Now we tackle solving the supremum in equation 9. First we consider the first-order condition on δ^- , namely that the partial derivative with respect to it must be equal to zero.

$$\begin{aligned}
0 &= \partial_{\delta^-} \left\{ (s + \pi + \delta^-) \mathbb{E}[L_k^-] - (s - \pi - \delta^+) \mathbb{E}[L_k^+] \right. \\
&\quad \left. + \mathbb{E}[L_k^+] (s + \mathbb{E}[\eta_{0,\mathbf{z}} T(\mathbf{z}, \omega)^{(2)}]) - \pi \mathbb{E}[L_k^+ \text{sgn}(q + L_k^+ - L_k^-)] \right. \\
&\quad \left. - \mathbb{E}[L_k^-] (s + \mathbb{E}[\eta_{0,\mathbf{z}} T(\mathbf{z}, \omega)^{(2)}]) + \pi \mathbb{E}[L_k^- \text{sgn}(q + L_k^+ - L_k^-)] \right. \\
&\quad \left. + q \mathbb{E}[\eta_{0,\mathbf{z}} T(\mathbf{z}, \omega)^{(2)}] - q \pi \mathbb{E}[\text{sgn}(q + L_k^+ - L_k^-)] + q \pi \text{sgn}(q) \right. \\
&\quad \left. + \mathbb{E}[h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^-)] - h_k(\mathbf{z}, q) \right\} \tag{31}
\end{aligned}$$

$$\begin{aligned}
&= \partial_{\delta^-} \left\{ (s + \pi + \delta^-) \mathbb{E}[L_k^-] - \pi \mathbb{E}[L_k^+ \text{sgn}(q + L_k^+ - L_k^-)] \right. \\
&\quad \left. - \mathbb{E}[L_k^-] (s + \mathbb{E}[\eta_{0,\mathbf{z}} T(\mathbf{z}, \omega)^{(2)}]) + \pi \mathbb{E}[L_k^- \text{sgn}(q + L_k^+ - L_k^-)] \right. \\
&\quad \left. - q \pi \mathbb{E}[\text{sgn}(q + L_k^+ - L_k^-)] + \mathbb{E}[h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^-)] \right\} \tag{32}
\end{aligned}$$

$$\begin{aligned}
&= p(\delta^-) - \kappa p(\delta^-) (s + \pi + \delta^-) - \pi \kappa p(\delta^+) p(\delta^-) \Psi_-(q) \\
&\quad + \kappa p(\delta^-) (s + \mathbb{E}[\eta_{0,\mathbf{z}} T(\mathbf{z}, \omega)^{(2)}]) - \pi \kappa p(\delta^-) \Upsilon(q, \delta^+) - q \pi \kappa p(\delta^-) \Phi_-(q, \delta^+) \\
&\quad + \kappa p(\delta^-) \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) [1 - 2p(\delta^+)] + h_{k+1}(\mathbf{j}, q-1) [-1 + p(\delta^+)] \right. \\
&\quad \left. + h_{k+1}(\mathbf{j}, q+1) [p(\delta^+)] \right] \tag{33}
\end{aligned}$$

Dividing through by $\kappa p(\delta^-)$, which is nonzero, and re-arranging, we find that the optimal sell

posting depth is given by

$$\begin{aligned} \delta^{-*} &= \frac{1}{\kappa} + \mathbb{E}[\eta_{0,z}T(\mathbf{z}, \omega)^{(2)}] - \pi \left(1 + p(\delta^+) \Psi_-(q) + \Upsilon(q, \delta^+) + q \Phi_-(q, \delta^+) \right) \\ &\quad + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) [1 - 2p(\delta^+)] + h_{k+1}(\mathbf{j}, q-1) [-1 + p(\delta^+)] + h_{k+1}(\mathbf{j}, q+1) [p(\delta^+)] \right] \end{aligned} \quad (34)$$

$$\begin{aligned} &= \frac{1}{\kappa} + \mathbb{E}[\eta_{0,z}T(\mathbf{z}, \omega)^{(2)}] - 2\pi \left(\mathbb{1}_{q \geq 1} + p(\delta^+) \mathbb{1}_{q=0} \right) \\ &\quad + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) [1 - 2p(\delta^+)] + h_{k+1}(\mathbf{j}, q-1) [-1 + p(\delta^+)] + h_{k+1}(\mathbf{j}, q+1) [p(\delta^+)] \right] \end{aligned} \quad (35)$$

$$\begin{aligned} \delta^{-*} &= \frac{1}{\kappa} + \mathbb{E}[\eta_{0,z}T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbb{1}_{q \geq 1} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) - h_{k+1}(\mathbf{j}, q-1) \right] \\ &\quad - p(\delta^+) \left(2\pi \mathbb{1}_{q=0} - \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q-1) + h_{k+1}(\mathbf{j}, q+1) - 2h_{k+1}(\mathbf{j}, q) \right] \right) \end{aligned} \quad (36)$$

And similarly, for the optimal buy posting depth:

$$\begin{aligned} \delta^{+*} &= \frac{1}{\kappa} - \mathbb{E}[\eta_{0,z}T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbb{1}_{q \leq -1} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) - h_{k+1}(\mathbf{j}, q+1) \right] \\ &\quad - p(\delta^-) \left(2\pi \mathbb{1}_{q=0} - \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q-1) + h_{k+1}(\mathbf{j}, q+1) - 2h_{k+1}(\mathbf{j}, q) \right] \right) \end{aligned} \quad (37)$$

For ease of notation we'll write $\aleph = \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} [h_{k+1}(\mathbf{j}, q-1) + h_{k+1}(\mathbf{j}, q+1) - 2h_{k+1}(\mathbf{j}, q)]$. Now, assuming we behave optimally on both the buy and sell sides simultaneously, we can substitute equation 37 into equation 36, and vice versa, while evaluating both at δ^{+*} and δ^{-*} to obtain the optimal posting depths in feedback form:

$$\begin{aligned} \delta^{-*} &= \frac{1}{\kappa} + \mathbb{E}[\eta_{0,z}T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbb{1}_{q \geq 1} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) - h_{k+1}(\mathbf{j}, q-1) \right] \\ &\quad - (1 - e^{\lambda(\mathbf{z})\Delta t}) e^{-\kappa \left(\frac{1}{\kappa} - \mathbb{E}[\eta_{0,z}T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbb{1}_{q \leq -1} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} [h_{k+1}(\mathbf{j}, q) - h_{k+1}(\mathbf{j}, q+1)] \right)} \\ &\quad \times e^{\kappa(1-e^{\lambda(\mathbf{z})\Delta t})e^{-\kappa\delta^{-*}}(2\pi \mathbb{1}_{q=0} - \aleph)} (2\pi \mathbb{1}_{q=0} - \aleph) \end{aligned} \quad (38)$$

$$\begin{aligned} \delta^{+*} &= \frac{1}{\kappa} - \mathbb{E}[\eta_{0,z}T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbb{1}_{q \leq -1} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) - h_{k+1}(\mathbf{j}, q+1) \right] \\ &\quad - (1 - e^{\lambda(\mathbf{z})\Delta t}) e^{-\kappa \left(\frac{1}{\kappa} + \mathbb{E}[\eta_{0,z}T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbb{1}_{q \geq 1} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} [h_{k+1}(\mathbf{j}, q) - h_{k+1}(\mathbf{j}, q-1)] \right)} \\ &\quad \times e^{\kappa(1-e^{\lambda(\mathbf{z})\Delta t})e^{-\kappa\delta^{+*}}(2\pi \mathbb{1}_{q=0} - \aleph)} (2\pi \mathbb{1}_{q=0} - \aleph) \end{aligned} \quad (39)$$

DPE Cases 2 and 3

We have derived the optimal LO posting depths for the first case of the DPE, where no market orders are placed. The analysis of the other two cases, where $M_k^+ = 1$ and $M_k^- = 1$, respectively, proceeds almost identically. Looking first at the case $M_k^+ = 1$, we try to solve:

$$\sup_{\delta^\pm} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ - (s + \pi) \right. \right. \\ \left. \left. + (L_k^+ - L_k^- + 1) (s + \eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^- + 1) \pi) \right. \right. \\ \left. \left. + q (\eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^- + 1) - \text{sgn}(q)) \pi) \right. \right. \\ \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^- + 1) - h_k(\mathbf{z}, q) \right] \right\}; \quad (40)$$

We find that the functions Φ, Ψ, Υ are identical but evaluated at $q + 1$. Thus, for example, we have:

$$\Phi(q + 1, \delta^+, \delta^-) = \begin{cases} 1 & q \geq 1 \\ 1 - p(\delta^-)(1 - p(\delta^+)) & q = 0 \\ p(\delta^+) - p(\delta^-) & q = -1 \\ -[1 - p(\delta^+)(1 - p(\delta^-))] & q = -2 \\ -1 & q \leq -3 \end{cases} \quad (41)$$

In a similar fashion to the previous analysis, we obtain the posting depths for the second case (indicated with a subscript 2)

$$\delta_2^{-*} = \frac{1}{\kappa} + \mathbb{E}[\eta_{0,z} T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbf{1}_{q \geq 0} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q + 1) - h_{k+1}(\mathbf{j}, q) \right] \\ - p(\delta^+) \left(2\pi \mathbf{1}_{q=-1} - \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) + h_{k+1}(\mathbf{j}, q + 2) - 2h_{k+1}(\mathbf{j}, q + 1) \right] \right) \quad (42)$$

$$\delta_2^{+*} = \frac{1}{\kappa} - \mathbb{E}[\eta_{0,z} T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbf{1}_{q \leq -2} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q + 1) - h_{k+1}(\mathbf{j}, q + 2) \right] \\ - p(\delta^-) \left(2\pi \mathbf{1}_{q=-1} - \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q) + h_{k+1}(\mathbf{j}, q + 2) - 2h_{k+1}(\mathbf{j}, q + 1) \right] \right) \quad (43)$$

Again, substitution equations 42 and 43 into one another and evaluating both at the optimal posting depths, we obtain

$$\delta_2^{-*} = \frac{1}{\kappa} + \mathbb{E}[\eta_{0,z} T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbf{1}_{q \geq 0} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} \left[h_{k+1}(\mathbf{j}, q + 1) - h_{k+1}(\mathbf{j}, q) \right] \\ - (1 - e^{\lambda(\mathbf{z})\Delta t}) e^{-\kappa \left(\frac{1}{\kappa} - \mathbb{E}[\eta_{0,z} T(\mathbf{z}, \omega)^{(2)}] - 2\pi \mathbf{1}_{q \leq -2} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z}, \mathbf{j}} [h_{k+1}(\mathbf{j}, q+1) - h_{k+1}(\mathbf{j}, q+2)] \right)} \\ \times e^{\kappa(1 - e^{\lambda(\mathbf{z})\Delta t}) e^{-\kappa \delta_2^{-*}}} (2\pi \mathbf{1}_{q=-1} - \aleph_{(+1)}) (2\pi \mathbf{1}_{q=-1} - \aleph_{(+1)}) \quad (44)$$

$$\begin{aligned}
\delta_2^{+*} = & \frac{1}{\kappa} - \mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z},\omega)^{(2)}] - 2\pi\mathbb{1}_{q\leq-2} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} \left[h_{k+1}(\mathbf{j}, q+1) - h_{k+1}(\mathbf{j}, q+2) \right] \\
& - (1 - e^{\lambda(\mathbf{z})\Delta t}) e^{-\kappa\left(\frac{1}{\kappa} + \mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z},\omega)^{(2)}] - 2\pi\mathbb{1}_{q\geq 0} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} [h_{k+1}(\mathbf{j}, q+1) - h_{k+1}(\mathbf{j}, q)]\right)} \\
& \times e^{\kappa(1-e^{\lambda(\mathbf{z})\Delta t})e^{-\kappa\delta_2^{+*}}} (2\pi\mathbb{1}_{q=-1-\aleph_{(+1)}}) (2\pi\mathbb{1}_{q=-1} - \aleph_{(+1)})
\end{aligned} \tag{45}$$

Likewise, in the third DPE case where $M_k^+ = 0, M_k^- = 1$, we get

$$\begin{aligned}
\delta_3^{-*} = & \frac{1}{\kappa} + \mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z},\omega)^{(2)}] - 2\pi\mathbb{1}_{q\geq 2} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} \left[h_{k+1}(\mathbf{j}, q-1) - h_{k+1}(\mathbf{j}, q-2) \right] \\
& - (1 - e^{\lambda(\mathbf{z})\Delta t}) e^{-\kappa\left(\frac{1}{\kappa} - \mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z},\omega)^{(2)}] - 2\pi\mathbb{1}_{q\leq 0} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} [h_{k+1}(\mathbf{j}, q-1) - h_{k+1}(\mathbf{j}, q)]\right)} \\
& \times e^{\kappa(1-e^{\lambda(\mathbf{z})\Delta t})e^{-\kappa\delta_3^{-*}}} (2\pi\mathbb{1}_{q=1-\aleph_{(-1)}}) (2\pi\mathbb{1}_{q=1} - \aleph_{(-1)})
\end{aligned} \tag{46}$$

$$\begin{aligned}
\delta_3^{+*} = & \frac{1}{\kappa} - \mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z},\omega)^{(2)}] - 2\pi\mathbb{1}_{q\leq 0} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} \left[h_{k+1}(\mathbf{j}, q-1) - h_{k+1}(\mathbf{j}, q) \right] \\
& - (1 - e^{\lambda(\mathbf{z})\Delta t}) e^{-\kappa\left(\frac{1}{\kappa} + \mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z},\omega)^{(2)}] - 2\pi\mathbb{1}_{q\geq 2} + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} [h_{k+1}(\mathbf{j}, q-1) - h_{k+1}(\mathbf{j}, q-2)]\right)} \\
& \times e^{\kappa(1-e^{\lambda(\mathbf{z})\Delta t})e^{-\kappa\delta_3^{+*}}} (2\pi\mathbb{1}_{q=1-\aleph_{(-1)}}) (2\pi\mathbb{1}_{q=1} - \aleph_{(-1)})
\end{aligned} \tag{47}$$

The boxed equations for optimal depth in feedback form will need to be solved numerically due to the difficulty in isolating $\delta^{\pm*}$ on one side of the equality.

Simplifying the DPE

For reference, we repeat here the DPE we are attempting to solve:

$$\begin{aligned}
0 = \max \Bigg\{ & \sup_{\delta^\pm} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ \right. \right. \\
& + (L_k^+ - L_k^-) (s + \eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^-) \pi) \\
& + q (\eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^-) - \text{sgn}(q)) \pi) \\
& \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^-) - h_k(\mathbf{z}, q) \right] \right\} ; \\
& \sup_{\delta^\pm} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ - (s + \pi) \right. \right. \\
& + (L_k^+ - L_k^- + 1) (s + \eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^- + 1) \pi) \\
& + q (\eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^- + 1) - \text{sgn}(q)) \pi) \\
& \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^- + 1) - h_k(\mathbf{z}, q) \right] \right\} ; \\
& \sup_{\delta^\pm} \left\{ \mathbb{E}_{\mathbf{w}} \left[(s + \pi + \delta^-) L_k^- - (s - \pi - \delta^+) L_k^+ + (s - \pi) \right. \right. \\
& + (L_k^+ - L_k^- - 1) (s + \eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - \text{sgn}(q + L_k^+ - L_k^- - 1) \pi) \\
& + q (\eta_{0,z} T(\mathbf{z}, \omega)^{(2)} - (\text{sgn}(q + L_k^+ - L_k^- - 1) - \text{sgn}(q)) \pi) \\
& \left. \left. + h_{k+1}(T(\mathbf{z}, \omega), q + L_k^+ - L_k^- - 1) - h_k(\mathbf{z}, q) \right] \right\} \Bigg\} \tag{48}
\end{aligned}$$

We now turn to simplifying our DPE by substituting in the optimal posting depths as written in recursive form (e.g. 37 and 36). In doing so we see a incredible amount of cancellation and

simplification, and we obtain the rather elegant, and surprisingly simple form of the DPE:

$$\begin{aligned}
h_k(\mathbf{z}, q) = \max \Bigg\{ & q\mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z}, \omega)^{(2)}] + \frac{1}{\kappa}(p(\delta^{+*}) + p(\delta^{-*})) - 2\pi p(\delta^{+*})p(\delta^{-*})\mathbf{1}_{q=0} \\
& + p(\delta^{+*})p(\delta^{-*}) \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} [h_{k+1}(\mathbf{j}, q-1) + h_{k+1}(\mathbf{j}, q+1) - 2h_{k+1}(\mathbf{j}, q)] \\
& + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} h_{k+1}(\mathbf{j}, q) ; \\
& (q+1)\mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z}, \omega)^{(2)}] + \frac{1}{\kappa}(p(\delta_2^{+*}) + p(\delta_2^{-*})) - 2\pi p(\delta_2^{+*})p(\delta_2^{-*})\mathbf{1}_{q=-1} \\
& + p(\delta_2^{+*})p(\delta_2^{-*}) \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} [h_{k+1}(\mathbf{j}, q) + h_{k+1}(\mathbf{j}, q+2) - 2h_{k+1}(\mathbf{j}, q+1)] \quad (49) \\
& + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} h_{k+1}(\mathbf{j}, q+1) ; \\
& (q-1)\mathbb{E}[\eta_{0,\mathbf{z}}T(\mathbf{z}, \omega)^{(2)}] + \frac{1}{\kappa}(p(\delta_3^{+*}) + p(\delta_3^{-*})) - 2\pi p(\delta_3^{+*})p(\delta_3^{-*})\mathbf{1}_{q=1} \\
& + p(\delta_3^{+*})p(\delta_3^{-*}) \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} [h_{k+1}(\mathbf{j}, q-2) + h_{k+1}(\mathbf{j}, q) - 2h_{k+1}(\mathbf{j}, q-1)] \\
& + \sum_{\mathbf{j}} \mathbf{P}_{\mathbf{z},\mathbf{j}} h_{k+1}(\mathbf{j}, q-1) \Bigg\}
\end{aligned}$$

TODO: commentary on this final form.

We now have an explicit means of numerically solving for the optimal posting depths. Since we know the function h at the terminal timestep T , we can take one step back to $T-1$ and solve for each of the optimal posting depths in each of the cases. With these values we are then able to calculate the value function h_{T-1} , and in doing so determine whether to execute market orders in addition to posting limit orders (essentially taking the $\arg \max$ instead of the \max in equation 49). This process then repeats for each step backward.