

You will find the code for this question in the `RL-book/Assignment2/assignment9_code.py` file.

CME 241: Assignment 9: Problem 2:

Let us derive the expressions for the optimal Value Function and Optimal Policy for the LPT price impact model.

$$\text{we have: } V_t^*(P_t, R_t) = \mathbb{E}_\pi \left( \sum_{i=t}^T N_i Q_i \mid P_t, R_t \right) = \mathbb{E}_\pi \left( \sum_{i=t}^T N_i (R_i (1 - \beta N_i) - \theta X_i) \mid P_t, R_t \right)$$

With the Bellman Equation, we get that:  $V_t^*(P_t, R_t) = \max_{N_t} \{ N_t Q_t + \mathbb{E}(V_{t+1}^*(P_{t+1}, R_{t+1})) \}$  assuming  $\gamma = 1$ 

$$\text{And } V_{T-1}^*(P_{T-1}, R_{T-1}) = N_{T-1} P_{T-1} (1 - \beta N_{T-1} - \theta X_{T-1}) = R_{T-1} Q_{T-1}$$

$$\text{From this, we can infer } V_{T-2}^*(P_{T-2}, R_{T-2}) = \max_{N_{T-2}} (N_{T-2} P_{T-2} (1 - \beta N_{T-2} - \theta X_{T-2}) + \mathbb{E}(R_{T-1} P_{T-1} (1 - \beta N_{T-1} - \theta X_{T-1}))$$

$$\text{Let us compute: } a = \mathbb{E}(R_{T-1} P_{T-1} (1 - \beta N_{T-1} - \theta X_{T-1}) \mid P_{T-2}, R_{T-2})$$

$$R_{T-1} = R_{T-2} - N_{T-2} \text{ and } P_{T-1} = P_{T-2} e^{N_{T-2}}, \text{ also } X_{T-1} = \rho X_{T-2} + \eta_{T-2}$$

$$\begin{aligned} \text{Hence } a &= \mathbb{E}((R_{T-2} - N_{T-2}) P_{T-2} e^{N_{T-2}} (1 - \beta R_{T-2} + \beta N_{T-2} - \theta \rho X_{T-2} + \theta \eta_{T-2})) \\ &= (R_{T-2} - N_{T-2}) P_{T-2} \left[ (1 - \beta R_{T-2} + \beta N_{T-2} - \theta \rho X_{T-2}) \mathbb{E}(e^{N_{T-2}}) + \theta \mathbb{E}(\eta_{T-2}) \right] \\ &\quad \exp\left(\frac{P_{T-2}^2}{2}\right) \exp\left(\frac{\rho^2 X_{T-2}^2}{2}\right) \times 0 \text{ by independence} \end{aligned}$$

This gives us:

$$a = (R_{T-2} - N_{T-2}) P_{T-2} (1 - \beta (R_{T-2} - N_{T-2}) - \theta \rho X_{T-2}) \exp\left(\frac{P_{T-2}^2}{2} + \frac{\rho^2 X_{T-2}^2}{2}\right)$$

$$\text{If we go back to: } V_{T-2}^*(P_{T-2}, R_{T-2}) = \max_{N_{T-2}} (N_{T-2} P_{T-2} (1 - \beta N_{T-2} - \theta X_{T-2}) + (R_{T-2} - N_{T-2}) P_{T-2} (1 - \beta (R_{T-2} - N_{T-2}) - \theta \rho X_{T-2}))$$

Let us derive this expression with respect to  $N_{T-2}$  and set it to 0

$$P_{T-2} (1 - \beta N_{T-2} - \theta X_{T-2}) - \beta N_{T-2} P_{T-2} - (R_{T-2} - N_{T-2}) P_{T-2} (1 - \beta (R_{T-2} - N_{T-2}) - \theta \rho X_{T-2}) e^{N_{T-2} + \frac{P_{T-2}^2}{2}} + \beta (R_{T-2} - N_{T-2}) P_{T-2} e^{N_{T-2} + \frac{P_{T-2}^2}{2}} = 0$$

$$1 - 2\beta N_{T-2} - \theta X_{T-2} (-1 + \beta R_{T-2} - \beta N_{T-2} - \theta \rho X_{T-2}) e^{N_{T-2} + \frac{P_{T-2}^2}{2}} + \beta R_{T-2} e^{N_{T-2} + \frac{P_{T-2}^2}{2}} - \beta N_{T-2} e^{N_{T-2} + \frac{P_{T-2}^2}{2}} = 0$$

$$1 - e^{N_{T-2} + \frac{P_{T-2}^2}{2}} + N_{T-2} (-2\beta - 2\beta e^{N_{T-2} + \frac{P_{T-2}^2}{2}}) + X_{T-2} (-\theta + \theta \rho e^{N_{T-2} + \frac{P_{T-2}^2}{2}}) + R_{T-2} (2\beta e^{N_{T-2} + \frac{P_{T-2}^2}{2}} - 1) = 0$$

This gives us that:

$$N_{T-2} = \frac{R_{T-2} e^{N_{T-2} + \frac{P_{T-2}^2}{2}}}{1 + e^{N_{T-2} + \frac{P_{T-2}^2}{2}}} + \frac{e^{N_{T-2} + \frac{P_{T-2}^2}{2}} - 1}{2\beta(1 + e^{N_{T-2} + \frac{P_{T-2}^2}{2}})} - \frac{\theta X_{T-2}}{2\beta} \times \frac{1 + e^{N_{T-2} + \frac{P_{T-2}^2}{2}}}{1 + e^{N_{T-2} + \frac{P_{T-2}^2}{2}}}$$

Continuing backwards gives us a similar solution for all  $t$  in  $[0, T]$ 

$$N_t = \frac{R_t e^{N_t + \frac{P_t^2}{2}}}{1 + e^{N_t + \frac{P_t^2}{2}}} + \frac{e^{N_t + \frac{P_t^2}{2}} - 1}{2\beta(1 + e^{N_t + \frac{P_t^2}{2}})} - \frac{\theta X_t}{2\beta} \times \frac{1 + e^{N_t + \frac{P_t^2}{2}}}{1 + e^{N_t + \frac{P_t^2}{2}}}$$

The solution is of the form:  $N_t = c_t^{(1)} + c_t^{(2)} R_t + c_t^{(3)} X_t$ , and plugging this in the optimal value function, we get:  $V^*(t, P_t, R_t, X_t) = e^{N_t + \frac{P_t^2}{2}} \cdot b_t \cdot (c_t^{(4)} + c_t^{(5)} R_t + c_t^{(6)} X_t + c_t^{(7)} R_t^2 + c_t^{(8)} X_t^2 + c_t^{(9)} R_t X_t)$ 

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