

Execution game in a Markovian environment

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1. Introduction

In a discrete time framework $t \in \{1, \dots, T, T+1\}$ ($T \in \mathbb{Z}_{++} := \{1, 2, \dots\}$), we assume that two large traders, denoted by $i \in \{1, 2\}$, purchase one risky asset in a trading market. Each large trader has a CARA vN-M (or negative exponential) utility function with the absolute risk aversion parameter $\gamma^i > 0$.

2. Model

Each large trader must purchase $\Omega^i (\in \mathbb{R})$ volume of one risky asset by the time $T+1$. $q_t^i (\in \mathbb{R})$ stands for large amount of orders of submitted by the large trader $i \in \{1, 2\}$ at time $t \in \{1, \dots, T\}$. We denote by \bar{Q}_t^i the number of shares remained to purchase by the large trader at time $t \in \{1, \dots, T, T+1\}$. So

$$\bar{Q}_{t+1}^i = \bar{Q}_t^i - q_t^i, \quad (1)$$

with the initial and terminal conditions: $\bar{Q}_1^i = \Omega^i \in \mathbb{R}$; $\bar{Q}_{T+1}^i = 0 \in \mathbb{R}$, for each large trader $i \in \{1, 2\}$.

The market price of the risky asset at time $t \in \{1, \dots, T, T+1\}$ is P_t . Then, the execution price of the asset becomes \hat{P}_t . We assume that submitting one unit of (large) order at time $t \in \{1, \dots, T\}$ causes the instantaneous price impact denoted as $\lambda_t (> 0)$.

We subsequently define the residual effect of past price impact caused by both large traders at time $t \in \{1, \dots, T\}$, represented by $R_t \in \mathbb{R}$. The dynamics of R_t for $t \in \{1, \dots, T\}$ is defined as:

$$R_{t+1} = e^{-\rho} [R_t + \alpha_t \lambda_t (q_t^1 + q_t^2)], \quad (2)$$

where $\alpha_t \in [0, 1]$ represents the linear price impact coefficients representing the temporary price impacts.

Furthermore, we define a sequence of independent random variables ϵ_t at time $t \in \{1, \dots, T\}$ as the effect of the public news/information about the economic situation between t and $t+1$. ϵ_t for $t \in \{1, \dots, T\}$ are assumed to follow a normal distribution with mean $\mu_t^\epsilon \in \mathbb{R}$ and variance $(\sigma_t^\epsilon)^2 \in \mathbb{R}_{++}$,

$$\epsilon_t \sim N\left(\mu_t^\epsilon, (\sigma_t^\epsilon)^2\right), \quad t = 1, \dots, T. \quad (3)$$

A *Markovian environment*, denoted by \mathcal{I}_t , influences the fundamental price. The distribution of \mathcal{I}_t is assumed to have a Markovian dependence:

$$\mathcal{I}_{t+1} | \mathcal{I}_t \sim N\left(a_{t+1}^{\mathcal{I}} - b_{t+1}^{\mathcal{I}} \mathcal{I}_t, (\sigma_{t+1}^{\mathcal{I}})^2\right). \quad (4)$$

Assumption 2.1. \mathcal{I}_t and ϵ_t are *correlated* for each $t \in \{1, \dots, T\}$. In addition, no other sequential dependence between two stochastic sequences exists.

By definition of ϵ_t , we define the dynamics of the fundamental price $P_t^f := P_t - R_t$ with a Markovian environment and the permanent price impact for time $t \in \{1, \dots, T\}$ as follows:

$$P_{t+1}^f := P_t^f + \beta_t \lambda_t (q_t^1 + q_t^2) + \mathcal{I}_t + \epsilon_t. \quad (5)$$

According to Eq. (2) and (5), the dynamics of market price for $t \in \{1, \dots, T\}$ are described as

$$P_{t+1} = P_t - (1 - e^{-\rho}) R_t + (\alpha_t e^{-\rho} + \beta_t) \lambda_t (q_t^1 + q_t^2) + \mathcal{I}_t + \epsilon_t. \quad (6)$$

The wealth process for each large trader $i \in \{1, 2\}$, $W_t^i (\in \mathbb{R})$, for $t \in \{1, \dots, T\}$ evolves as follows:

$$W_{t+1}^i = W_t^i - \hat{P}_t q_t^i = W_t^i - \{P_t + \lambda_t (q_t^1 + q_t^2)\} q_t^i. \quad (7)$$

3. Formulation as a Markov game

We define the state of the decision process at time $t \in \{1, \dots, T, T+1\}$ as 7-tuple and denote it as

$$\mathbf{s}_t = \left(W_t^1, W_t^2, P_t, \bar{Q}_t^1, \bar{Q}_t^2, R_t, \mathcal{I}_{t-1}\right) \in \mathbb{R}^7 =: S. \quad (8)$$

For $t \in \{1, \dots, T\}$, an allowable action chosen at state \mathbf{s}_t is an execution volume $q_t^i \in \mathbb{R} =: A^i$ so that the set A^i of admissible actions is independent of the current state \mathbf{s}_t . When an action q_t^i is chosen in a state \mathbf{s}_t at time $t \in \{1, \dots, T\}$, a transition to a next state: $\mathbf{s}_{t+1} = (W_{t+1}^1, W_{t+1}^2, P_{t+1}, \bar{Q}_{t+1}^1, \bar{Q}_{t+1}^2, R_{t+1}, \mathcal{I}_t) \in S$ occurs according to the law of motion which we have precisely described in the previous subsection.

A utility payoff (or reward) arises only in a terminal state s_{T+1} at the end of horizon $T + 1$ as

$$g_{T+1}^i(s_{T+1}) := \begin{cases} -\exp\{-\gamma^i W_{T+1}^i\} & \text{if } \bar{Q}_{T+1}^i = 0; \\ -\infty & \text{if } \bar{Q}_{T+1}^i \neq 0, \end{cases} \quad (9)$$

where $\gamma^i > 0$ represents the risk aversion parameter for the large trader $i \in \{1, 2\}$. The term $-\infty$ means a hard constraint enforcing the large trader to execute all of the remaining volume \bar{Q}_T^i at the maturity T . The types of large traders could be defined by $(W^i, \Omega^i, \gamma^i)$, $i \in \{1, 2\}$, and these are assumed to be their common knowledge.

If we define a (history-independent) one-stage decision rule f_t at time $t \in \{1, \dots, T\}$ by a Borel measurable map from a state $s_t \in S = \mathbb{R}^7$ to an action

$$q_t^i = f_t^i(s_t) \in A = \mathbb{R}, \quad (10)$$

then a Markov execution strategy π is defined as a sequence of one-stage decision rules

$$\pi^i := (f_1^i, \dots, f_t^i, \dots, f_T^i). \quad (11)$$

We denote the set of all Markov execution strategies as Π_M . Further, for $t \in \{1, \dots, T\}$, we define the sub-execution strategy after time t of a Markov execution strategy $\pi \in \Pi_M$ as

$$\pi_t^i := (f_t^i, \dots, f_T^i), \quad (12)$$

and the entire set of π_t^i as $\Pi_{M,t}^i$.

By definition (9), the value function under an execution strategy profile (π^1, π^2) becomes an expected utility payoff arising from the terminal wealth W_{T+1} :

$$V_1^i(\pi^1, \pi^2)[s_1] = \mathbb{E}_1^{\pi^1, \pi^2} \left[g_{T+1}^i(s_{T+1}) \mid s_1 \right], \quad (13)$$

where, for $t \in \{1, \dots, T\}$, $\mathbb{E}_t^{\pi^1, \pi^2}$ is a conditional expectation given a condition at time t under the strategy profile (π^1, π^2) .

For $t \in \{1, \dots, T, T+1\}$ and $s_t \in S$, we let

$$V_t^i(\pi_t^1, \pi_t^2)[s_t] = \mathbb{E}_t^{\pi_t^1, \pi_t^2} \left[g_{T+1}^i(s_{T+1}) \mid s_t \right], \quad (14)$$

be the expected utility payoff at time t .

Definition 3.1 (Nash Equilibrium). $(\pi^1, \pi^2) \in \Pi_M^1 \times \Pi_M^2$ is a Nash equilibrium starting from a fixed initial state s_1 if and only if

$$V_1^1(\pi^1, \pi^2)[s_1] \geq V_1^1(\pi^1, \pi^{2*})[s_1], \quad \forall \pi^1 \in \Pi_M^1; \quad (15)$$

$$V_1^2(\pi^1, \pi^2)[s_1] \geq V_1^2(\pi^{1*}, \pi^2)[s_1], \quad \forall \pi^2 \in \Pi_M^2. \quad (16)$$

Definition 3.2 (Markov Perfect Equilibrium). $(\pi^1, \pi^2) \in \Pi_M^1 \times \Pi_M^2$ is a Markov perfect equilibrium if and only if for all $t \in \{1, \dots, T\}$ and for all $s_t \in S$,

$$V_t^1(\pi_t^1, \pi_t^{2*})[s_t] \geq V_t^1(\pi_t^1, \pi_t^2)[s_t], \quad \forall \pi_t^1 \in \Pi_{M,t}^1, \quad (17)$$

$$V_t^2(\pi_t^{1*}, \pi_t^2)[s_t] \geq V_t^2(\pi_t^1, \pi_t^2)[s_t], \quad \forall \pi_t^2 \in \Pi_{M,t}^2. \quad (18)$$

Based on the One Stage [Step, Shot] Deviation Principle, we obtain an equilibrium execution strategy at a Markov perfect equilibrium by backward induction procedure of dynamic programming.

4. Equilibrium execution strategy

Theorem 4.1. There exists a Markov perfect equilibrium at which the following properties hold for each large trader $i \in \{1, 2\}$:

1. The execution volume at the Markov perfect equilibrium for the large trader $i \in \{1, 2\}$ at time $t \in \{1, \dots, T\}$, denoted as q_t^{i*} , becomes an *affine function* of $(\bar{Q}_t^i, \bar{Q}_t^j, \mathcal{I}_{t-1})$:

$$q_t^{i*} = a_t^i + b_t^i \bar{Q}_t^i + c_t^i \bar{Q}_t^j + d_t^i R_t + e_t^i \mathcal{I}_{t-1}. \quad (19)$$

2. The value function $V_t^i(\pi^1, \pi^2)[s_t]$ at time $t \in \{1, \dots, T, T+1\}$ for each large trader $i \in \{1, 2\}$ is represented as a functional form:

$$\begin{aligned} V_t^i(\pi_t^1, \pi_t^2)[s_t] = & -\exp \left\{ -\gamma \left[W_t^i - P_t^\top \bar{Q}_t^i + G_t^{1i} \left(\bar{Q}_t^i \right)^2 \right. \right. \\ & + G_t^{2i} \bar{Q}_t^i + H_t^{1i} \bar{Q}_t^i R_t + H_t^{2i} R_t^2 + H_t^{3i} R_t + I_t^{1i} \bar{Q}_t^i \bar{Q}_t^j \\ & + I_t^{2i} \bar{Q}_t^j R_t + I_T^{3i} \left(\bar{Q}_t^j \right)^2 + I_t^{4i} \bar{Q}_t^j + J_t^{1i} \bar{Q}_t^i \mathcal{I}_{t-1} + J_t^{2i} R_t \mathcal{I}_{t-1} \\ & \left. \left. + J_t^{3i} \bar{Q}_t^j \mathcal{I}_{t-1} + J_t^{4i} \mathcal{I}_{t-1}^2 + J_t^{5i} \mathcal{I}_{t-1} + Z_t^i \right] \right\}. \quad (20) \end{aligned}$$

The dynamics of $a_t^i, b_t^i, c_t^i, d_t^i, e_t^i$; $G_t^{1i}, G_t^{2i}, H_t^{1i}, H_t^{2i}, H_t^{3i}, I_t^{1i}, I_t^{2i}, I_t^{3i}, I_t^{4i}, J_t^{1i}, J_t^{2i}, J_t^{3i}, J_t^{4i}, J_t^{5i}, Z_t^i$ for $t \in \{1, \dots, T, T+1\}$ are deterministic functions of time t which are dependent on the problem parameters, and can be computed backwardly in time t from maturity T .

参考文献

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