

# Asset Pricing under Asymmetric Information Strategic Market Order Models

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# A Classification of Market Microstructure Models

- simultaneous submission of demand schedules
  - competitive rational expectation models
  - strategic share auctions
- sequential move models
  - screening models in which the market maker submits a supply schedule first
    - static
      - ◇ uniform price setting
      - ◇ limit order book analysis
    - dynamic sequential trade models with multiple trading rounds
  - strategic market order models where the market maker sets prices ex-post

# Strategic Market Order Models - Overview

- Kyle (1985) model
  - static version
  - dynamic version (in discrete time)
    - Refresher in Dynamic Programming
  - continuous time version (Back 1992)
- Multi-insider Kyle (1985) version
- Other strategic market order models

## Kyle 1985 Model

- Model Setup

- asset return  $v \sim \mathcal{N}(\mu_0, \Sigma_0)$
- Agents (risk neutral)
  - Insider who knows  $v$  and submit market order of size  $x$
  - Noise trader who submit market orders of exogenous aggregate size  $u \sim \mathcal{N}(0, \sigma_u^2)$
  - Market maker sets competitive price after observing **net** order flow  $X = x + u$
- Timing (order of moves)
  - Stage 1: Insider & liquidity traders submit market orders
  - Stage 2: Market Maker sets the execution price
- Repeated trading in dynamic version

# Kyle 1985 Model — Static Version

## Single informed trader

### 0) Information

$v$  := asset's payoff

### 1) Conjecture (price-rule)

$$p = \mu + \lambda(x + u)$$

### 2) No Updating

### 3) Optimal Demand

$$\max_x E[(v - p)|v]x$$

$$\max_x E[v - \mu - \lambda x | v]x$$

$$\text{FOC: } x = -\frac{\mu}{2\lambda} + \frac{1}{2\lambda}v$$

$$\text{SOC: } \lambda > 0$$

### 4) Correct Beliefs

$$\alpha = -\frac{\mu}{2\lambda}, \beta = \frac{1}{2\lambda}$$

## (Competitive) Market Maker

### 0) Information

$X = x + u$  batch net order flow

### 1) Conjecture (insider trading rule)

$$x = \alpha + \beta v$$

### 2) Updating $E[v|x + u]$

### 3) Price Setting Rule

$$p = E[v|x + u]$$

$$p = E[v] + \frac{\text{Cov}[v, x+u]}{\text{Var}[x+u]} \{x + u - E[x + u]\}$$

$$p = p_0 + \frac{\beta \Sigma_0}{\beta^2 \Sigma_0 + \sigma_u^2} \{x + u - \alpha + \beta E[v]\}$$

### 4) Correct Beliefs

$$\mu = p_0 \text{ Martingale, } \lambda = \frac{\beta \Sigma_0}{\beta^2 \Sigma_0 + \sigma_u^2}$$

## Kyle 1985 Model — Static Version

- solve for unknown coefficients
  - 4 unknown Greeks
  - 4 equations
- $\lambda = \frac{1}{2} \sqrt{\frac{\Sigma_0}{\sigma_u^2}}$ 
  - $\lambda$  (illiquidity) decreases with noise trading,  $\sigma_u^2$
  - $\Sigma_0$  reflect info advantage of insider

## Dynamic Programming - A Refresher

- The Problem:  
max for several periods  $t = 1, \dots, T$  (discrete time)

$$\max_{u_t} E_t \left[ \sum_{s=1}^T v_s (\tilde{x}_s, u_s) \right] \quad \forall t \Rightarrow (\text{sequential rationality})$$

under the following law of motion

$$\tilde{x}_{t+1} = f_t(\tilde{x}_t, u_t, \tilde{\varepsilon}_t)$$

$\tilde{x}_t$ : vector of state variables (sufficient state space)

$u_t$ : vector of control variables

$\varepsilon_t$ : vector of random shocks

- Method
  - Backward Induction
  - Dynamic Programming

## Dynamic Programming - A Refresher

- Define **Value Function**

$$V_t(x_t) := E_t \left[ \sum_{s=t}^T v_s(\tilde{x}_s, \underset{\substack{\uparrow \\ \text{optimal values}}}{u_s^*}) \right]$$

- $\Rightarrow$  **Bellman Equation**

$$\max_{u_t} E_t [v_t(x_t, u_t) + V_{t+1}(x_{t+1})]$$

- Start at final **date**  $T$

$$V_{T+1}(\cdot) := 0$$

$\Rightarrow$  in  $t = T$

$$\max_{u_T} E_T [v_T(x_T, u_T)]$$

$$\stackrel{\text{FOC}}{\Rightarrow} u_T^* = g_T(x_T)$$

$$\Rightarrow V_T(x_T) = E_T [v_T(x_T, u_T^*)]$$

## Dynamic Programming - A Refresher

- at **date**  $T - 1$

$$\max_{u_{T-1}, u_T} E_{T-1} \left[ \sum_{s=T-1}^T v_s (\tilde{x}_s, u_s) \right]$$

given  $V_T(x_T)$

$$\Leftrightarrow \max_{u_{T-1}} E_{T-1} [v_{T-1}(x_{T-1}, u_{T-1}) + V_T(x_T)]$$

given law of motion

$\Leftrightarrow$

$$\max_{u_{T-1}} E_{T-1} [v_{T-1}(x_{T-1}, u_{T-1}) + V_T(f_{T-1}(x_{T-1}, u_{T-1}, \tilde{\varepsilon}_{T-1}))]$$

$$\Rightarrow u_{T-1}^* = g_{T-1}(x_{T-1})$$

$$\Rightarrow V_{T-1} =$$

$$E_{T-1} [v_{T-1}(x_{T-1}, u_{T-1}^*) + V_T(f_{T-1}(x_{T-1}, u_{T-1}^*, \tilde{\varepsilon}_{T-1}))]$$

- and so on for **date**  $T - 2$  etc. (and if they didn't die in the uncertainties they are still solving ...)
- This process is quite time consuming.

# Dynamic Programming - A Refresher

Alternative way:

- **Step 1:** “Guess” the general form of the value function

$$V_{t+1}(x_{t+1}) = \underbrace{H_{t+1}(x_{t+1})}_{\text{e.g. } H_{t+1}(x_{t+1}) = \alpha_{t+1}x_{t+1}^2}$$

- **Step 2:** Derive optimal level of current control

$$\begin{aligned} \max_{u_t} E_t [v_t(x_t, u_t) + H_{t+1}(\tilde{x}_{t+1})] \\ \max_{u_t} E_t [v_t(x_t, u_t) + H_{t+1}(f_t(x_t, u_t, \varepsilon_t))] \\ \Rightarrow u_t^* = \dots \end{aligned}$$

- **Step 3:** Derive value function and check whether it coincides with general value function

$$\begin{aligned} V_t(x_t) &= E_t [v_t(x_t, u_t^*) + H_{t+1}(f_t(x_t, u_t^*, \tilde{\varepsilon}_t))] \\ \stackrel{?}{=} H_t(x_t) &= \alpha_t x_t^2 \end{aligned}$$

## Kyle (1985) — Dynamic Version

### Insider

- **Step 1:** Conjectured price setting strategy (pricing rule)

$$\begin{aligned} p_n &= p_{n-1} + \lambda_n \Delta X_n \\ &= p_{n-1} + \lambda_n (\Delta x_n + \Delta u_n) \quad \left( \frac{1}{\lambda_t} \cong \text{Liquidity} \right) \end{aligned}$$

- **Step 2:** 'Guess' Value function for insider's profit pricing rule is linear  $\rightarrow$  guess quadratic value function)

$$E[\pi_{n+1} | \underbrace{\tilde{p}_1, \dots, p_n, v}_{\text{Information set up to } n}] = \alpha_n (v - p_n)^2 + \delta_n$$

(expected profit from time  $n + 1$  onwards)

$$\pi_n = E_n [\pi_{n+1} + (v - p_n) \Delta x_n^i]$$

## Kyle (1985) — Dynamic Version

Insider ctd.

- **Step 3:** Write Bellman Equation

$$\max_{\Delta x_n^i} E \left[ (v - p_n) \Delta x_n^i + \alpha_n (v - p_n)^2 + \delta_n \underbrace{|p_1, \dots, p_{n-1}, v}_{I_{n-1}} \right]$$

- **Step 4:** Given insider's beliefs  $p_n = p_{n-1} + \lambda_n \Delta X_n$

$$\max_{\Delta x_n^i} E \left[ (v - p_{n-1} - \lambda_n \Delta x_n^i - \lambda_n \Delta u_n) \Delta x_n^i + \alpha_n (v - p_{n-1} - \lambda_n \Delta x_n^i - \lambda_n \Delta u_n)^2 + \delta_n \mid I_n \right]$$

Take expectations

$$\max_{\Delta x_n^i} \left[ (v - p_{n-1} - \lambda_n \Delta x_n^i) \Delta x_n^i + \alpha_n \underbrace{(v - p_{n-1} - \lambda_n \Delta x_n^i)^2}_{\text{control}} + \delta_n + \alpha_n \lambda_n^2 \underbrace{\sigma_u^2}_{\substack{\text{state variable} \\ u \Rightarrow p_n \text{ noisy}}} \Delta t_n \right]$$

## Kyle (1985) — Dynamic Version

Insider ctd.

- **Step 5:** maximize

$$\text{FOC: } (v - p_{n-1}) - 2\lambda_n \Delta x_n^i - 2\alpha_n \lambda_n (v - p_{n-1}) + 2\alpha_n \lambda_n^2 \Delta x_n^i = 0$$

$$\Delta x_n^i = \underbrace{\frac{1 - 2\alpha_n \lambda_n}{2\lambda_n (1 - \alpha_n \lambda_n)}}_{:= \beta_n \Delta t_n} (v - p_{n-1})$$

$$\text{SOC: } \lambda_n (1 - \alpha_n \lambda_n) > 0$$

- **Step 6:** Check whether 'guessed' value fcn is correct

$$E[\pi | I_{n-1}] = \max_{\Delta x_n^i} E \left[ (v - p_n) \Delta x_n^i + \alpha_n (\tilde{v} - \tilde{p}_n)^2 + \delta_n | I_{n-1} \right]$$

$$= \alpha_{n-1} (v - p_{n-1})^2 + \delta_{n-1}, \text{ where}$$

$$\alpha_{n-1} = \frac{1}{4\lambda_n (1 - \alpha_n \lambda_n)}, \quad \delta_{n-1} = \delta_n + \alpha_n (\lambda_n)^2 \sigma_u^2 \Delta t_n$$

## Kyle (1985) — Dynamic Version

### Market Maker (Filtering Problem)

- **Step 1:** MM's belief about insider's strategy

$$\Delta x_n^i = \beta_n \Delta t_n (v - p_{n-1})$$

$$\Delta X_n = \beta_n \Delta t_n (v - p_{n-1}) + \underbrace{\Delta u_n}_{\text{Var}[\Delta u_n] = \sigma_u^2 \Delta t_n}$$

- **Step 2:** MM's filtering problem

By definition:

$$p_{n-1} : = E[v | \Delta X_1, \dots, \Delta X_{n-1}]$$

$$\Sigma_{n-1} : = \text{Var}[v | \Delta X_1, \dots, \Delta X_{n-1}]$$

$$E[\Delta X_n | \Delta X_1, \dots, \Delta X_{n-1}] = \beta_n \Delta t_n E[(v - p_{n-1}) + \Delta u_n | \dots]$$

$$\text{Var}[\Delta X_n | \dots] = (\beta_n \Delta t_n)^2 \Sigma_{n-1} + \sigma_u^2 \Delta t_n$$

$$\text{Cov}[v, \Delta X_n | \dots] = E[v(\beta_n \Delta t_n (v - p_{n-1}) + \Delta u_n | \dots)]$$

$$= \beta_n \Delta t_n \Sigma_{n-1}$$

## Kyle (1985) — Dynamic Version

Now we have all ingredients for the **Projection Theorem**

$$p_n = p_{n-1} + \underbrace{\frac{\beta_n \Delta t_n \Sigma_{n-1}}{(\beta_n \Delta t_n)^2 \Sigma_{n-1} + \Delta t \sigma_u^2}}_{:=\lambda_n} \Delta X_n$$

$$\Sigma_n = V[v | \dots \Delta X_n] = \Sigma_{n-1} - \frac{(\beta_n \Delta t_n)^2 \Sigma_{n-1}^2}{(\beta_n \Delta t_n)^2 \Sigma_{n-1} + \Delta t \sigma_u^2}$$

$$= \frac{\sigma_u^2 \Sigma_{n-1}}{(\beta_n)^2 \Delta t_n \Sigma_{n-1} + \sigma_u^2}$$

$\Rightarrow$

$$\lambda_n = \frac{\beta_n \Delta t_n \Sigma_{n-1}}{(\beta_n \Delta t_n)^2 \Sigma_{n-1} + \Delta t \sigma_u^2}$$

$$\Sigma_n = (1 - \lambda_n \beta_n \Delta t_n) \Sigma_{n-1} = \frac{\sigma_u^2 \lambda_n}{\beta_n}$$

$$\Rightarrow \lambda_n = \beta_n \Sigma_n / \sigma_u^2$$

## Kyle (1985) — Dynamic Version

- **Step 3:** Equate coefficients  $\alpha_n, \beta_n, \delta_n, \Sigma_n$

$$\left. \begin{aligned} \beta_n \Delta t_n &= \frac{1 - 2\alpha_n \lambda_n}{2\lambda_n(1 - \alpha_n \lambda_n)} \\ \alpha_{n-1} &= \frac{1}{4\lambda_n(1 - \alpha_n \lambda_n)} \\ \delta_{n-1} &= \delta_n + \alpha_n \lambda_n^2 \sigma_u^2 \Delta t_n \\ \Sigma_n &= \sigma_u^2 \Sigma_{n-1} \\ \lambda_n &= \dots \end{aligned} \right\} \begin{array}{l} \text{Solve recursive} \\ \text{system of} \\ \text{equations.} \end{array}$$

} as above

- Interpretation of Equilibrium
  - restrain from aggressive trading
    - price impact in current trading round
    - price impact in all future trading rounds
  - ...

## Generalizations of Kyle (1985)

- Multiple Insiders
  - all have same information
  - all hold different information
  - information is correlated
    - ⇒ see Foster & Viswanathan, JF 51, 1437-1478
- Risk averse insiders
  - CARA utility
- etc. etc.