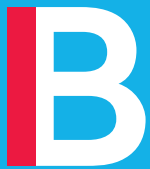


## Lecture 3b

### Price dynamics and liquidity - II



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### Outline

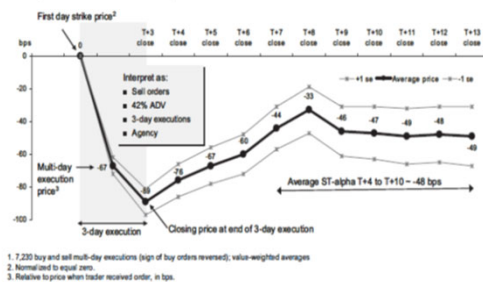
- Bid-ask spread and price dynamics with:
  - order processing costs
  - rents due to imperfectly competitive dealers
  - inventory holding costs: risk-averse dealers
- Putting it all together: contributions to price dynamics of
  - (i) adverse selection
  - (ii) order processing costs and rents
  - (iii) inventory risk

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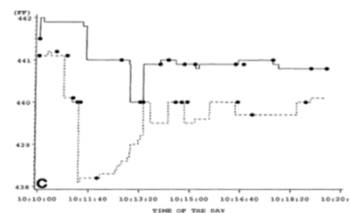
## Stylized fact

- Trades in financial markets have both a permanent and transient effects on prices
- Remember the return reversals in Kraus and Stoll (1972). Same in the graphs on the right, drawn from more recent studies
- Is informed order flow sufficient to explain this? No!

EXHIBIT 6  
Multi-day Executions & Short-term Alpha<sup>3</sup>



Source: Cai and Sofianos (2006)



Source: Biais et al. (1995)

## Price impact of orders under asymmetric information

- To see this, recall that in the previous lecture we showed that under asymmetric information the price is given by:

$$p_t = \mu_t = E(v | \Omega_{t-1}, d_t) = \mu_{t-1} + s(d_t)d_t$$

- So, under those assumptions, the price is a random walk, and the short-run impact  $s(d_t)d_t$  is expected to persist permanently:

$$p_t = E \left[ E(p_{t+\tau} | \Omega_{t+\tau-1}, d_{t+\tau}) | \Omega_{t-1}, d_t \right] = E(p_{t+\tau} | \Omega_{t-1}, d_t)$$

- Hence, with asymmetric information alone, orders do not generate mean reversion in prices

## Return reversals

- So, what is missing from the model with informed order flow? Can we enrich it so as to capture price reversals?
- We can, if we allow for
  - **order processing costs** (due to real costs of executing trades)
  - **inventory holding costs** (due to risk borne by market makers)
- This is what we turn to now

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## Order processing costs

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## Order processing costs

- Processing an order requires time and money:
  - exchange, clearing and settlement fees
  - cost of paperwork and back-office work, etc.
- Some of these order processing costs (OPC) are
  - on a per-share (or per-dollar traded) basis: variable cost
  - on a per-transaction basis: fixed cost
- Here we assume *variable* OPC:  $\gamma$  per share
  - but many results generalize to per-dollar and per-transaction OPC

## Bid-ask spread with order processing costs

- Suppose that orders are informative *and* OPC are  $\gamma$  per share
- The bid-ask spread must compensate dealers both (i) for their losses to informed traders and (ii) for the OPC:

$$\left. \begin{aligned} a_t &= \mu_{t-1} + \gamma + s_t^a \\ b_t &= \mu_{t-1} - \gamma - s_t^b \end{aligned} \right\} \Rightarrow S_t = \underbrace{2\gamma}_{\text{OPC}} + \underbrace{s_t^a + s_t^b}_{\text{adverse selection}}$$

- Can we measure the relative importance of these two components of the spread?

## Price dynamics with order processing costs

- From these ask and bid prices, the transaction price is:

$$p_t = \mu_{t-1} + s(d_t)d_t + \gamma d_t$$

- Since the expectation of the stock's fundamental is still:

$$\mu_t = \mu_{t-1} + s(d_t)d_t,$$

the transaction price deviates from the stock's fundamental value by the size of the OPC:

$$p_t = \mu_t + \gamma d_t$$

## Short-run price impact of a buy order

- How much does the transaction price deviate from the *midquote* (not the fundamental) after a buy order?

- The midquote is

$$m_t = \frac{a_t + b_t}{2} = \mu_{t-1} + \frac{s_t^a - s_t^b}{2}$$

- So the short-term impact of the buy order is

$$p_t - m_t = \overbrace{\mu_{t-1} + s_t^a}^{a_t} + \gamma - m_t = \frac{s_t^a + s_t^b}{2} + \gamma > 0$$

## Long-run price impact of a buy order

- What is the effect of the same order after  $T$  periods?
- The expected price is

$$E(p_{t+T}) = \underbrace{\mu_{t-1} + s_t^a}_{\mu_t^+} + \underbrace{\gamma E(d_{t+T} | \Omega_{t-1}, d_t)}_0$$

- So the long-term impact is expected to be:

$$E(p_{t+T}) - m_t = \underbrace{\mu_{t-1} + s_t^a}_{\mu_t^+} - m_t = \frac{s_t^a + s_t^b}{2}$$

in the LR, the OPC component  $\gamma$  vanishes: only the informational one persists

## Rents or order processing costs?

- $\gamma$  may not just capture costs, but also oligopoly profits (per share traded) accruing to dealers:

$$a_t = \mu_{t-1} + s_t^a + \underbrace{\gamma^p}_{\text{rent per share}} + \underbrace{\gamma^c}_{\text{OPC}}$$

- Same on the bid side
- So, the OPC component of the bid-ask spread may also include non-competitive rents!

## Inventory holding costs

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## Inventory holding costs

- Assume no informed traders ( $\pi = 0$ ) and no OPC ( $\gamma = 0$ ): still, a bid-ask spread emerges as compensation for inventory risk
- After filling orders, dealers end up with unbalanced portfolios (“inventories”): short after buys, long after sells
- If risk-averse, they want to be paid to bear inventory risk (“holding costs”)  $\Rightarrow$  bid-ask spread
- Dealer’s wealth after supplying  $y_t$  shares out of inventory  $z_t$ :

$$\begin{aligned}
 w_{t+1} &= \overbrace{p_{t+1} z_{t+1}}^{\text{final inventory}} + \overbrace{c_{t+1}}^{\text{final cash}} \\
 &= \underbrace{p_{t+1}}_v \underbrace{(z_t - y_t)}_{z_{t+1}} + \underbrace{c_t + p_t y_t}_{c_{t+1}} = v z_t + c_t + \underbrace{(p_t - v) y_t}_{\text{profits}}
 \end{aligned}$$

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## Static model with risk-averse dealer

- Consider a 2-period model:  $t = 0$  and  $t + 1 = 1$
- Suppose the dealer has a mean-variance objective function:

$$U = E(w_1) - \frac{\rho}{2} \text{var}(w_1)$$

- His objective as a function of the sale  $y_0$  is

$$\begin{aligned} U(y_0) &= E(v)z_0 + c_0 + [p_0 - E(v)]y_0 - \frac{\rho}{2}(z_0 - y_0)^2 \text{var}(v) \\ &= \mu_0 z_0 + c_0 + \underbrace{(p_0 - \mu_0)y_0}_{\text{expected profits}} - \frac{\rho}{2} \underbrace{(z_0 - y_0)^2 \sigma_\varepsilon^2}_{\text{inventory risk}} \end{aligned}$$

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## Dealer's supply function and price quotes

- The first-order condition yields the dealer's inverse supply function (price required to supply  $y_0$  shares):

$$p_0 = \mu_0 + \rho \sigma_\varepsilon^2 (y_0 - z_0)$$

- In equilibrium, the dealer's supply  $y_0$  equals investors' order  $q_0$ :

$$p_0 = \underbrace{\mu_0 - \rho \sigma_\varepsilon^2 z_0}_{m_0} + \rho \sigma_\varepsilon^2 q_0 = \begin{cases} a_0 = m_0 + \rho \sigma_\varepsilon^2 |q_0| & \text{if } q_0 > 0 \\ b_0 = m_0 - \rho \sigma_\varepsilon^2 |q_0| & \text{if } q_0 < 0 \end{cases}$$

- Bid-ask spread:

$$S_t = 2\rho\sigma_\varepsilon^2 |q_0| \Rightarrow \text{increasing in}$$

- (i) dealers' **risk aversion**
- (ii) fundamental **volatility**
- (iii) trade **size**

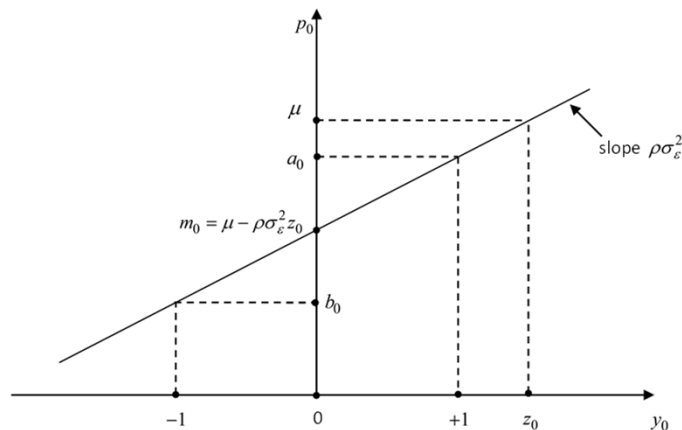
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## Dealer's quotes for a unit buy or sell order : $|q_t| = 1$



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## Price pressure

- The midquote  $m_0$  differs from the expected value  $\mu_0$  as it is the dealer's marginal valuation of the asset given his current inventory  $z_0$ :

$$m_0 = \mu_0 - \rho \sigma_\varepsilon^2 z_0$$

- Intuition:**

- if the dealer is long ( $z_0 > 0$ ), his marginal valuation is lower than  $\mu_0$  because extra shares add to his risk exposure
- if he is short ( $z_0 < 0$ ), his marginal valuation is higher than  $\mu_0$  because extra shares lower his risk exposure

⇒ **price pressure:** orders affect the dealers' midquote because they affect his inventory, hence his marginal valuation ⇒ over time, **midquotes and inventories are mirror images** of each other

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### Three main testable implications of inventory model

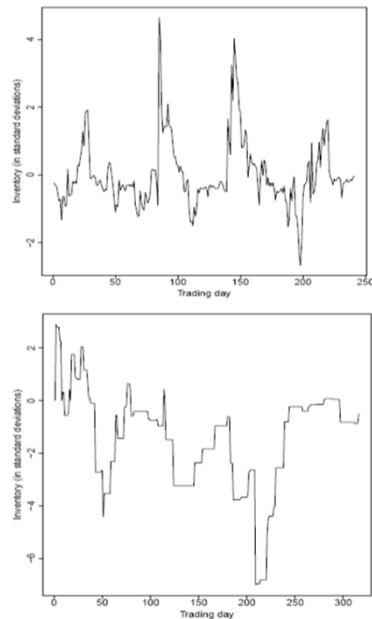
1. The price pressure of an order is inversely related to the inventory imbalance produced by the order: dealers with long (short) positions are willing to trade at lower (higher) prices
2. Receiving offsetting orders helps dealers unwind their positions  $\Rightarrow$  accelerate inventory reversion to target
3. Prices are also mean-reverting, at a speed that reflects that of reversion in inventories (mirror images)

### 1. Price pressure is inversely related to inventory position

- For stocks listed on the NYSE, Hendershott and Menkveld (2014) estimate the price pressure per unit of inventory
- They find that the higher dealers' inventories, the lower the midquote: their relationship features a negative coefficient
- Larger in absolute value for small-cap stocks: a \$100,000 inventory results in 1.01% price pressure for small stocks and 0.02% for large stocks
- As there is less trading in small stocks, it takes longer to unwind inventories and dealers remain exposed to risk for longer: they find that price pressure lasts longer for small-caps

## 2. Offsetting orders allow mean reversion in inventories

- Daily inventory of the dollar position of a block market making desk of a major broker-dealer for the Apple stock (source: Duffie, 2012)
- Daily inventory of a market making desk of a major broker-dealer for a single investment-grade corporate bond (source: Duffie, 2012)



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## What determines the speed of mean reversion?

- Mean reversion is faster when
  - the market is more active (e.g., large-cap vs. small cap, Apple stock vs. corporate bonds): less than 1 day for liquid stocks vs. 2 months for some small stocks
  - dealers manage to elicit inventory-reducing orders: Reiss and Werner (1998) and Hansch, Naik, and Viswanathan (1998) find that dealers with long positions are more likely to execute buy market orders; those with short positions, are more likely to execute sell market orders
  - there are no regulatory limits on the size of dealers' positions, such as short-sales constraints or margin constraints on leverage

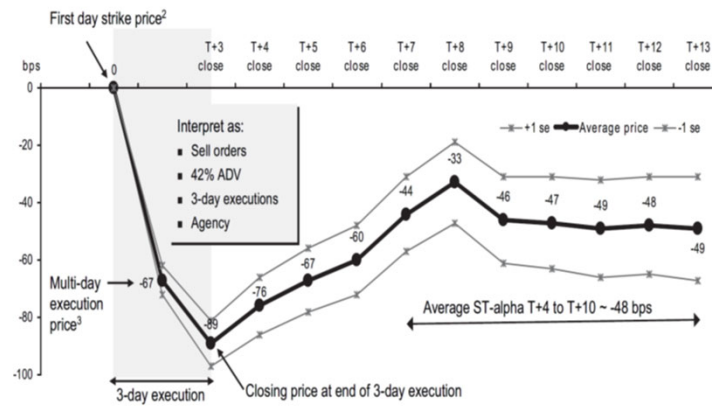
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### 3. Mean reversion in prices mirrors that in inventories

- Mean reversion in prices is especially strong for large stock orders executed over 3 days (from Cai and Sofianos, 2006, seen at start of this class)



- Mean reversion in prices is faster in more active markets, in sync with the greater speed of mean reversion of inventories: price pressure for NYSE stocks lasts longer for small-cap stocks, whose trading is less frequent (Hendershott and Menkveld, 2014)

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### The full picture

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## The full picture: effects of a buy order

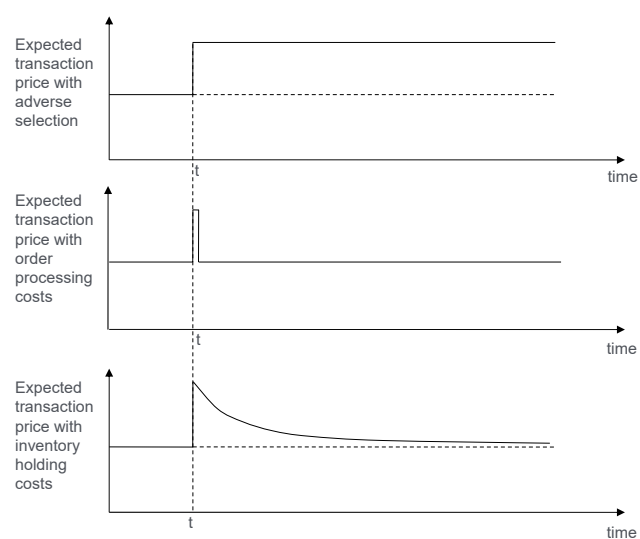
- **Adverse selection:**
  - Short run price impact = long run price impact  $> 0$
- **Order processing costs:**
  - Short run price impact  $> 0$
  - Long and medium run price impact = 0
- **Inventory holding costs:**
  - Short run price impact  $> 0$
  - Medium run price impact  $> 0$  but declining
  - Long run price impact = 0

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## Graphically: separately

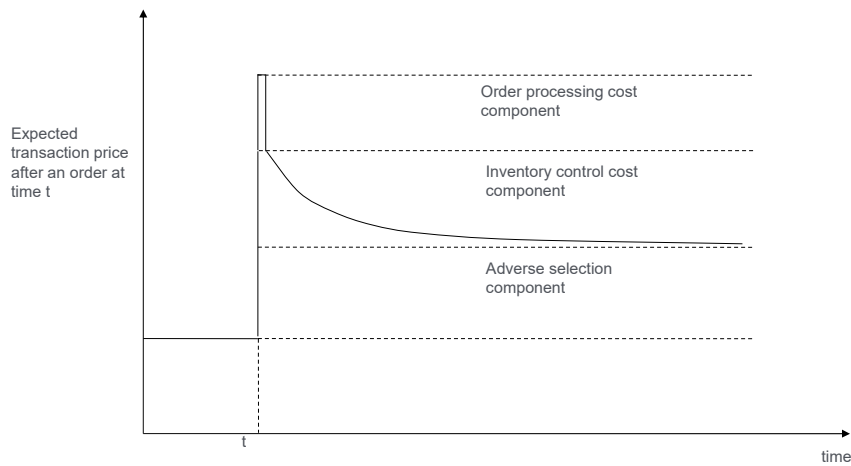


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## Graphically: all together



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## Key takeaways

- The response of asset prices to orders often features some reversion to the initial level, which is inconsistent with models of informed trading
- Order-processing costs and inventory holding costs can account for such patterns, the first at high frequency, the second at lower frequencies
- Inventory holding costs also account for the fact that
  - midquotes are negatively related to dealers' inventories
  - their respective dynamics after large orders are "mirror images" of each other

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