

Price Discovery in High Resolution

by

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Discussion

by

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Quick summary

General Goal

Study the **incorporation of information** into prices across markets and timescales

Framework

Hasbrouck's (1995) pioneering work who introduces:

- **VECM** for prices of the same asset formed in different markets
- **Information Share (IS)** as measure of price discovery of different markets

Main issue still open in the literature: allocation of correlation terms → **IS bounds**

Approach of the paper

- Shrinking IS bounds by **increasing resolution of obs up to microseconds**,
- Problem: Huge n° of lags in VECM ($\sim 10^6$)
- Solution: Retain parsimony with "extreme" versions of HAR structure

Main results

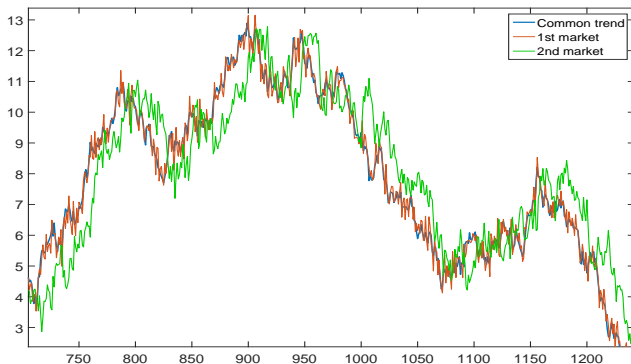
- Successfully **collapsing the IS bounds** as resolution increases
- Strong information dominance of **direct subscribers** over consolidated tape
- Slight information dominance of **listing exchanges** over other exchanges
- Strong information dominance of **quotes** over trades and **lit** over dark trades

Ad Hoc Simulation I: VECM with one Lag

- We simulate a VECM for two market prices with 1 lag:

$$\Delta p_t = \gamma B p_{t-1} + \phi_1 \Delta p_{t-1} + \epsilon_t, \quad \epsilon_t \sim N(0, \Omega), \quad \Omega \text{ is diagonal}$$

- With a choice of ϕ_1 that induces a **sizable lag for the 2nd market**



IS 1st market = [0.486, 0.483],

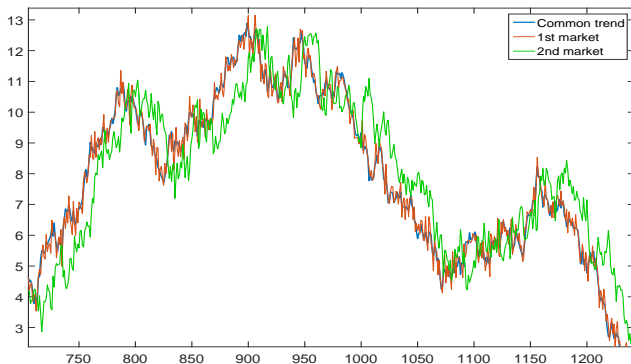
IS 2nd market = [0.513, 0.516]

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IS and Autoregressive Lags ϕ_i

- Using the common trends representation

$$p_t = \theta(1)_* \left(\sum_{s=1}^t \epsilon_s \right) \iota + \theta^*(L)\epsilon_t \equiv m_t \iota + s_t$$

with $\theta(1)_*$ a generic row of $\theta(1)$ and ι a vector of ones.

- The **Information Share (IS)** is defined as

$$IS_i = \frac{\theta(1)_{*,i}^2 \Omega_{i,i}}{\theta(1)_* \Omega \theta(1)'_*}$$

with $\theta(1)_{*,i}$ the i -th element of $\theta(1)_*$

- But, from the Granger's representation theorem we can analytically compute

$$\theta(1) = B_{\perp} \underbrace{\left[\gamma'_{\perp} \left(I_N - \sum_{i=1}^k \phi_i \right) B_{\perp} \right]^{-1}}_{\substack{(N-r) \times (N-r) \\ \text{if } r=N-1 \Rightarrow \text{scalar}}} \gamma'_{\perp}$$

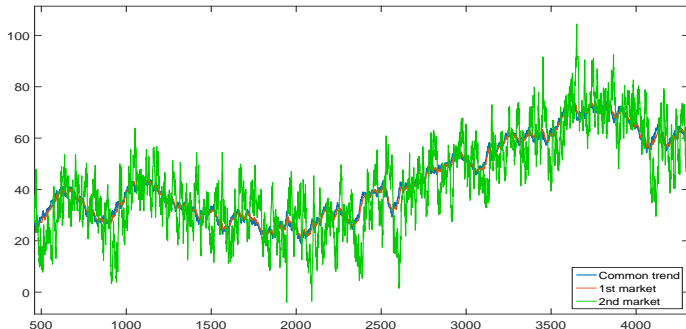
- Hence, the term containing the autoregressive lags ϕ_i **cancel out** in the IS ratio
 \Rightarrow **IS depends on γ and Ω , but not DIRECTLY on ϕ 's.**

Ad Hoc Simulation II: VECM with Different Variances

- We simulate a VECM for two markets (with no lags):

$$\Delta p_t = \gamma B p_{t-1} + \epsilon_t, \quad \epsilon_t \sim N(0, \Omega), \quad \Omega \text{ is diagonal}$$

- **1st market** is fast adapting (γ_1 small) and with small innovation variance
- **2nd market** is slower (γ_2 larger) and with substantial innovation variance



IS 1st market = [0.380, 0.389],

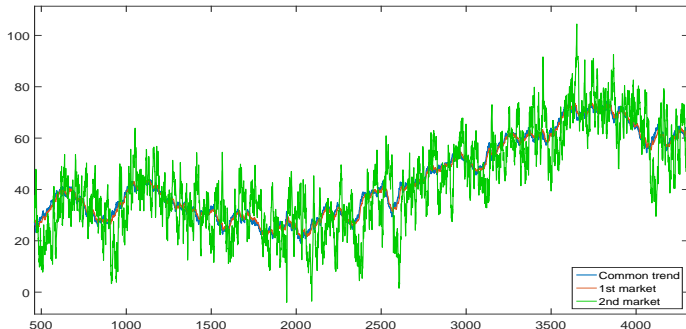
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IS and Market Variances $\Omega_{i,i}$

- In the simulations with a **reduced form VECM**, the IS of a market depends:
 - Positively on the speed of adjustment of the market:
 $\text{small } \gamma_i \Rightarrow \text{high } IS_i$
 - Positively on the noise variance of the market:
 $\text{large } \Omega_{i,i} \Rightarrow \text{high } IS_i$
- \Rightarrow IS of a mkt with sufficiently large $\Omega_{i,i}$ can be larger than a mkt with a small γ_j (**Update**: see also the presentation here at SoFiE pre-conf of **Christian Nguenang**)
- Not clear if the reduced form model VECM contains MM noise or not (see **Dias, Fernandes, Scherrer 2018**)
- Difficulties of interpretation might stem from the reduced form nature of VECM
- Look for a structural model (as in De Jong & Schotman 2010) but with a lagged adjustment mechanism
- We consider a **multivariate generalization of the partial price adjustment** model of Hasbrouck & Ho (1987)

Structural Model with Lagged Adjustments

- In the proposed model the **efficient price** m_t is the scalar process

$$m_t = m_{t-1} + u_t, \quad u_t \sim \text{NID}(0, \Sigma)$$

- **The latent prices with lagged adjustment** X_t is assumed to be

$$X_t = X_{t-1} + \Psi(m_t - X_{t-1})$$

the matrix Ψ is the **speed of adjustment matrix**.

If $\Psi = \mathbb{I}$, then $X_{i,t} = m_t$, i.e. instantaneous price adjustment in every market

- Finally, the **vector of observed log-prices** is:

$$P_t = X_t + \varepsilon_t, \quad \varepsilon_t \sim \text{NID}(0, H)$$

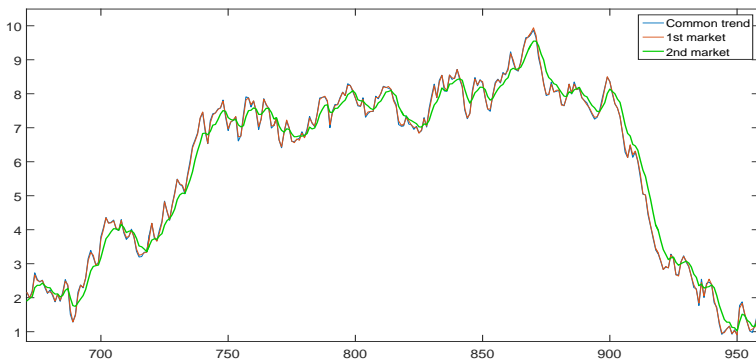
- This model can be conveniently estimated with the standard Kalman filter (treating zeros as missing values) assuming (identification assumption) $u_t \perp \varepsilon_t$

Structural Model with Partial Price Adjustment

- We simulate the structural model for two markets:

$$P_t = X_{t-1} + \Psi(m_t \iota - X_{t-1}) + \varepsilon_t, \quad \varepsilon_t \sim N(0, \Omega), \quad \Omega \text{ is diagonal}$$

- **1st market** is fast adapting ($\Psi_{1,1} = 0.9$)
- **2nd market** is slower ($\Psi_{2,2} = 0.3$)



IS 1st market = [0.997, 0.087]

IS 2nd market = [0.912, 0.002]

Concluding Remarks

Some additional questions:

- What are the effects of large number of zero returns on IS?
- What is the effect of the strong one-second periodicity observed in UHF data?

What I liked about the paper:

- Clean and effective solution to the problem of non-uniqueness of IS
- Heterogeneous/Multifrequency view of the market
- Unified approach to modeling price dynamics across a full range of frequencies
- Consistency of results across resolutions
- Emergence of clearer pictures as resolution increases.