

Liquidity Dynamics and Stock Market Automation

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October 19, 2008





Motivation - Effects of Automation on Liquidity

- 1 Trade-offs between lower operating costs and higher adverse selection costs.
- 2 Trade-offs between execution costs, quantities and speeds.
- 3 More information may allow discretionary traders to cluster, resulting in greater liquidity spikes and droughts. Admati & Pfleiderer [1988], Coppejans Domowitz & Madhavan [2001]



A Selective Summary of Findings in the Literature

Single Exchange Event Studies - Mixed Evidence

Lower costs: Israel & France - Amihud, Mendelson & Lauterbach [1997], Kalay, Wei & Wohl [2002], Muscarella & Piwowar

Higher costs: US & Germany - Theissen [1999] & Henderschott Moulton [2008]

Cross-sectional Studies - Mixed Samples

Hard to compare one exchange to another

Lower costs: 51 exchanges - Jain [2005]

Higher costs: NYSE v. Paris - Venkatamaran [2001]

Liquidity Dynamics - No Evidence

Persistence, Volatility and Cross-moments of Liquidity (AP Liquidity Betas)



Data

Daily data available 49 countries

Bid-Ask spreads are typically not available until after automation.

Countries Excluded from the Event Study

Reason

Number

Countries Affected

(1) Not automated by 2005

3

Ghana, Kenya and Zimbabwe

(2) Price availability coincides with Automation

11

Brazil, Bulgaria, Chile, China, Estonia, Jamaica
Jordan, Latvia, Luxembourg, Romania and Saudi Arabia

(3) Price availability more than 1 year after Automation

9

Iceland, Kuwait, Lithuania, Nigeria, Oman,
Russia, Slovakia, Taiwan and Tunisia



Plan

- 1 Pre/post analysis with Bootstrap Errors
- 2 Speed of Information Incorporation
- 3 Firm-level Panel Regressions
- 4 Persistence and Volatility of Systematic Liquidity
- 5 Acharya Pedersen [2005] Liquidity CAPM (with GMM Structural Break Tests)



Liquidity Proxies

LOT - Lesmond Ogden Trzcinka [1999]

ZEROS - As used in Bekaert, Harvey Lundblad [2007]

$$ZEROS = T^{-1} \sum_{t=1}^T I_{(R_t=0)}$$

ILLIQ - The Amihud [2002] illiquidity measure

$$ILLIQ = T^{-1} \sum_{t=1}^T \frac{|R_t|}{VO_{dol,t}}, \quad VO_{dol,t} > 0$$

LCF is a latent common factor representation of **LOT**, **ZEROS** and, where available, **ILLIQ**.



Liquidity Before and After Automation

Panel A. All Countries (49 countries)				
Statistic	LOT	ZEROS	ILLIQ	LCF
Pre-Automation	5.7	34.0	11.1	-0.39
Post-Automation	2.7	25.5	10.0	-0.68
Proportion Falling	71%	76%	82%	86%
Average Change	-2.6*	-8.1***	-1.0***	-0.27***
Wilcoxon Mann-Whitney	(0.040)	(0.002)	(0.045)	(0.001)
Bootstrap P-value	(0.066)	(0.001)	(0.002)	(0.003)
Median Change	-0.51*	-5.1***	-0.77**	-0.19**
Bootstrap P-value	(0.079)	(0.005)	(0.037)	(0.011)

Robustness: Results similar at 6 months, 1 year, 2 years and 4 year windows.



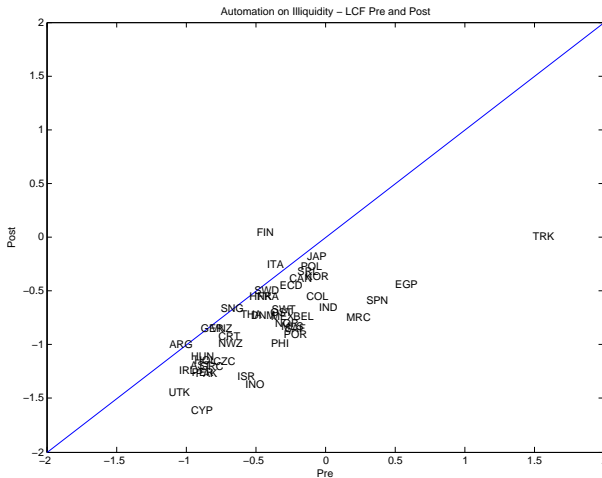
Liquidity Before and After Automation by Development

Panel B. Developed Countries (20 countries)				
Statistic	LOT	ZEROS	ILLIQ	LCF
Pre-Automation	2.2	26.5	10.5	-0.44
Post-Automation	2.2	24.2	9.7	-0.58
Proportion Falling	60%	75%	93%	80%
Average Change	0.00	-2.3	-0.9	-0.14
Median Change	-0.2	-2.7	-0.6	-0.14

Panel C. Emerging Markets (29 countries)				
Statistic	LOT	ZEROS	ILLIQ	LCF
Pre-Automation	8.3	39.5	11.4	-0.35
Post-Automation	3.1	26.5	10.3	-0.75
Proportion Falling	79%	76%	72%	90%
Average Change	-4.4	-12.0	-1.1	-0.36
Median Change	-0.8	-11.8	-1.0	-0.28



Automation on Illiquidity: Country-by-Country Results (LCF)



Speed of Information Transmission

Delay Measures from Mech [1993], Hou & Moskowitz [2005]

$$Delay = \max(0, \bar{R}_{SW}^2 - \bar{R}_{MM}^2)$$

where \bar{R}_{SW}^2 is the adjusted R^2 from a Scholes-Williams [1977]/Dimson [1979] type regression and \bar{R}_{MM}^2 is the adjusted R^2 from the typical market model regression.

We also compute variance ratios, $VR(q)$ at $q = \{2, 5, 10\}$ trading days for individual securities and for market indices,

$$VR(q) = 1 + 2 \sum_{k=1}^{q-1} \left(1 - \frac{k}{q}\right) \hat{\rho}(k),$$

following Campbell, Lo & MacKinlay [1997].



Changes in Speed of Information Incorporation

For Individual Securities.

All Countries (49 countries)

Statistic	Delay	VR(2)	VR(5)	VR(10)
Pre-Automation	0.055	1.001	0.896	0.688
Post-Automation	0.051	0.987	0.857	0.648
Proportion Falling	65%	69%	78%	76%
Average Change	-0.004	-0.015*	-0.043***	-0.044***
Wilcoxon Mann-Whitney	(0.226)	(0.034)	(0.006)	(0.019)
Bootstrap P-value	(0.510)	(0.064)	(0.003)	(0.002)
Median Change	-0.006**	-0.016**	-0.039***	-0.036***
Bootstrap P-value	(0.035)	(0.046)	(0.002)	(0.004)

Small-Large Portfolio Cross-correlations will be in the next draft.



Changes in Speed of Information Incorporation by Development

Developed Countries (20 countries)				
Statistic	Delay	VR(2)	VR(5)	VR(10)
Pre-Automation	0.057	0.995	0.884	0.671
Post-Automation	0.052	0.975	0.846	0.640
Proportion Falling	80%	80%	75%	75%
Average Change	-0.002	-0.019	-0.037	-0.024
Median Change	-0.006	-0.019	-0.035	-0.028

Emerging Markets (29 countries)				
Statistic	Delay	VR(2)	VR(5)	VR(10)
Pre-Automation	0.053	1.004	0.905	0.701
Post-Automation	0.051	0.996	0.865	0.655
Proportion Falling	55%	62%	79%	76%
Average Change	-0.001	-0.019	-0.031	-0.038
Median Change	-0.006	-0.012	-0.041	-0.049



Systematic Liquidity Persistence and Volatility

The persistence of systematic liquidity are from the autoregression

$$LCF_{*W,t} = \psi + \rho LCF_{*W,t-1} + \varepsilon_t$$

where $LCF_{*W,t}$ is the illiquidity proxy aggregated up by $*W \in \{EW, VW\}$ method in time, t . ρ is the autoregressive coefficient.

$$\sigma_L = \sqrt{T^{-1} \sum_{t=1}^T e_t^2}$$

where T is the total number of time series observations and e_t is the estimate residual at time t from above.



Persistence & Volatility of Systematic Liquidity

All Countries (49 countries)				
Statistic	<i>PLCF,EW</i>	<i>PLCF,VW</i>	$\sigma_{LCF,EW}$	$\sigma_{LCF,VW}$
Average	0.21	0.16	-35.4	-19.4
Median	0.19	0.10	-39.6	-25.7
Proportion Falling	24%	33%	88%	74%
Wilcoxon Mann-Whitney	0.00	0.00	0.00	0.00

Results stronger in emerging markets than in developed ones. To do: Structural break tests...



Liquidity Risks

Acharya-Pedersen [2005] liquidity-CAPM that captures the following covariances

- 1 Sharpe-Lintner Beta, β_{MKT} , Asset Returns and Market Returns (Market Beta)
- 2 Chordia-Roll-Subrahmanyam Beta, β_{CRS} , Asset Liquidity and Market Liquidity (Commonality)
- 3 Pastor-Stambaugh Beta, β_{PS} , Asset Returns and Market Liquidity
- 4 Acharya-Pedersen Beta, β_{AP} , Asset Liquidity and Market Returns

$$COV_{MKT,*W} = cov(R_{i,t}, R_{t,*W} - E_{t-1}R_{t,*W}),$$

$$COV_{CRS,*W} = cov(L_{i,t} - E_{t-1}L_{i,t}, L_{t,*W} - E_{t-1}L_{t,*W}),$$

$$COV_{PS,*W} = cov(R_{i,t}, L_{t,*W} - E_{t-1}L_{t,*W}),$$

$$COV_{AP,*W} = cov(L_{i,t} - E_{t-1}L_{i,t}, R_{t,*W} - E_{t-1}R_{t,*W}),$$



Liquidity Risks

All Countries (39 countries)

Statistic	COV_{MKT}			COV_{CRS}		
	PRE	POST	DIFF	PRE	POST	DIFF
Average	0.90	0.55	-0.35	0.58	0.37	-0.21
Median	0.62	0.41	-0.18	0.47	0.29	-0.17
Proportion Falling			77%			82%
Wilcoxon Mann-Whitney			0.01			0.01

Statistic	COV_{PS}			COV_{AP}		
	PRE	POST	DIFF	PRE	POST	DIFF
Average	-0.58	-0.36	0.22	-0.34	-0.27	0.07
Median	-0.39	-0.16	0.09	-0.37	-0.12	0.10
Proportion Falling			41%			28%
Wilcoxon Mann-Whitney			0.16			0.07



Summary of Findings and Extensions

- 1 Automation appears to reduce transactions costs by 3% over the whole sample
 - Mixed results in the literature may be dependent on the initial liquidity of the market. Developed, efficient floor-trading markets may not have much room for improvement.
 - Automation may make emerging markets more accessible to foreign traders, improving liquidity but opening the market up to global liquidity crises.
- 2 Automation is consistent with more frequent trading and faster incorporation of information
- 3 Automation appears to reduce liquidity risk
 - Are liquidity crises a peso problem?
- 4 A few more things to do

