Are Market Makers Incentivized to Provide Liquidity? Evidence from the Nasdag

Nikolaus Hautsch¹ Julia Reynolds²

¹University of Vienna

²Università della Svizzera italiana

August 19, 2019

3rd SAFE Market Microstructure Conference

Incentive Schemes and the Role of Liquidity Provision

- Rise of "endogenous liquidity providers" (ELPs) instead of traditional markets makers has led many equity trading platforms to implement liquidity incentive schemes.
- Questions remain as to the primary role of a liquidity provider:
 - Patient Counterparties: Provision of limit orders (Grossman and Miller, 1988).
 - Contrarian Traders: Price stabilization (Hendershott and Menkveld, 2014;
 Van Kervel and Menkveld, 2019).
- Most incentive programs focus on the provision of limit orders, but contrarian trading may be essential to mitigating textcolorbluemarket fragility!
 (Bessembinder et al., 2015; Anand and Venkataraman, 2016; Biais et al., 2017)
- Can an incentive programs be successful if it fails to promote both roles?

Our Paper

Research Questions

- How effective are liquidity incentive programs in promoting liquidity provision in modern markets?
 - To what extent do the qualifying orders submitted by market makers serve as (1) patient counterparties, or (2) contrarian traders?
- How does the market react to non-anonymous orders posted by market makers?
- Focus on Nasdaq's Qualified Market Maker (QMM) program, which mandates a certain quota of limit orders per month, because:
 - Choice of when and under what circumstances to provide liquidity at market maker's discretion.
 - Qualifying quotes required to be non-anonymous.
 - Purpose of program was to attract liquidity from non-traditional agents.

Our Approach

- Unique dataset reconstructs limit order book for several Nasdaq-traded stocks, including market participant identifier numbers (MPIDs).
- Drivers of Market Maker Submissions: Fixed effects panel regression of market market MPID submission intensities on lagged market characteristics.
- Market Reactions to Market Maker Submissions: Fixed effects panel regression of market quality measures on lagged market maker MPID submission intensities.
 - Address endogeneity concerns using the Heckman (1979) correction

Drivers of MM Submissions: Hypotheses

- H1 If market makers are incentivized to provide liquidity, then we should see an increase in their submissions in response to:
 - Lower submission volume and higher execution volumes.
 - Lower depth.
 - Higher volatility (Chung and Chuwonganant, 2014).
 - Higher bid-ask spreads (Chordia et al., 2008).
- H1a If market makers act in the capacity as "contrarian traders," then we should additionally see an increase in their submissions:
 - In the opposite direction of prices changes (Hendershott and Menkveld, 2014).
 - In response to an increase in pricing errors (Hasbrouck, 1993).

Market Reactions to MM Submissions: Hypotheses

- H2 If the MPID-attributed orders are submitted by market markers in their capacity as liquidity providers, then in response to higher rates of MPID submission we should see:
 - Higher execution and submission volume.
 - Higher depth.
- H2a If MPID-attributed orders succeed in attracting reactive (v. parasitic) counterparties, then we should *additionally* see:
 - Lower bid-ask spreads and lower volatility (Harris, 1997).
- H2b If market makers act in the capacity as "contrarian traders," then we should additionally see:
 - Decrease in pricing errors (Watanabe, 2017; Anand and Venkataraman, 2016).

Nasdaq Qualified Market Maker (QMM) Program

- Introduced in 2012 and designates a market participant as a Qualified Market Maker (QMM) if they:
 - Maintained quotes at the national best bid or offer (NBBO),
 - In at least 1,000 securities,
 - During at least 25% of trading hours.
 - Qualifications assessed on a monthly basis.
- In return, QMMs receive credits for submissions and executions, and along with reduced liquidity take fees.
- QMMs need not be registered market makers.
- QMM Program "designed to attract liquidity both from traditional market makers and from other firms that are willing to commit capital to support liquidity at the NBBO." (SEC Release No. 34-70361, 10 September 2013)

Data

- Reconstructed Nasdaq limit order book obtained from LOBSTER Academic Data.
 - Order book data on prevailing bid and ask quotes and depths.
 - Message file updates (submissions, partial or total cancellations, and executions of visible or hidden orders).
- Sample composed of eight Nasdaq-listed stocks, mostly in the high-tech industry.
- Sample time period from 4-22 November 2013.
- Data uniquely contains information on the Market Participant Identification Number (MPID) of non-anonymous orders.

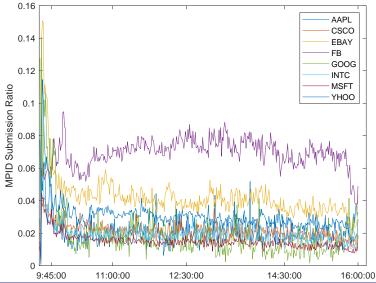
MPID	Firm Name	MPID Type	%Sub.
ATDF	Automated Trading Desk Financial Services, LLC	Market Maker	0.05%
BARD	Robert W. Baird & Co. Incorporated	Market Maker	< 0.01%
DADA	D.A. Davidson & Co.	Market Maker	< 0.01%
FBCO	Credit Suisse Securities (USA) LLC	Market Maker	0.28%
GSCO	Goldman, Sachs & Co.	Market Maker	1.55%
RHCO	Suntrust Robinson Humphrey, Inc.	Market Maker	< 0.01%
SBSH	Citigroup Global Markets Inc.	Market Maker	17.21%
TMBR	Timber Hill LLC	Market Maker	76.36%
UBSS	UBS Securities LLC	Market Maker	4.44%
WCHV	Wells Fargo Securities, LLC.	Market Maker	< 0.01%
		Total Market Maker	99.89%
воок	Bloomberg Tradebook LLC	ECN	0.03%
LEHM	Barclays Capital Inc./Le	Nasdag Participant	0.01%
NITE	Knight Capital Americas LLC	Nasdaq Participant	0.03%
WEMM	Wells Fargo Securities, LLC.	Nasdag Participant	0.04%
	, , , , , , , , , , , , , , , , , , ,	Total Other	0.11%

MPID Submission Intensities

- Decision to submit an MPID-attributed order \approx decision to submit a QMM-qualifying order.
- MPID Submission intensities defined for stock i in interval [t, t+1] as:

$$MPID_t^i = \frac{\#MPIDSubmissions_t^i}{\#TotalSubmissions_t^i}.$$

- Focus on number of submissions isolates submission decision from other order choices (i.e., the size and price of the order).
- Aggregate over 60-second intervals.



■ Fixed effects panel regression for stock i in interval [t, t+1]:

$$\mathit{MPID}_t^i = \alpha_0^i + \theta' \sum_{p=1}^{20} \mathit{MPID}_{t-p}^i + \beta' \mathbf{x}_t^i + \gamma' \mathbf{m}_{t-1}^i + \delta' \mathit{DAY}_t + \varepsilon_t^i,$$

- \mathbf{x}_t^i : average characteristics of MPID-attributed orders.
- \mathbf{m}_{t-1}^{i} : market conditions.
- DAY_t , α_0^i : day and stock fixed effects.
- $MPID_{t-p}^{i}$: to account for high autocorrelation.
- Order Characteristics: order aggressiveness and size.
- Market Conditions: relative bid-ask spreads, volatility, submission and execution volumes, depth, (unsigned) change in price, negative/positive price change dummy, pricing error (Hasbrouck, 1993), dummy capturing market open.

Dep.Var	(1) $MPID_t^i$	(2) MPID ^{i;BUY}	(3) MPID ^{i;BUY}	(4) MPID ^{i;SELL}	(5) MPID ^{i;SELL}
\mathbb{I}_t^{OPEN}	0.139***	0.154***	0.147***	0.129***	0.132***
-1	(5.907)	(6.646)	(6.393)	(5.302)	(5.349)
Rel. Bid-Ask Spreads	0.0631***	0.0504***	0.0521***	0.0441***	0.0440***
•	(7.575)	(6.462)	(6.682)	(5.682)	(5.735)
Volatility	0.0744***	0.0654***	0.0637***	0.0589***	0.0620***
	(6.474)	(6.599)	(6.523)	(4.696)	(5.119)
Observations	43,520	43,520	43,520	43,520	43,520
Number of Stocks	8	8	8	8	8
Stock FE	YES	YES	YES	YES	YES
Day FE	YES	YES	YES	YES	YES
Lagged Dep. Var	YES	YES	YES	YES	YES
Within R ²	0.352	0.283	0.283	0.264	0.264

- Market markers submit MPID orders when spreads and volatility are high.
- ⇒ Supports Hypothesis H1.

	(1)	(2)	(3)	(4)	(5)
Dep.Var	MPID!	MPID _t ;BUY	MPID _t ;BUY	MPID _t ;SELL	MPID _t ;SELL
Submissions	-0.0513***				
	(-6.146)				
Executions	0.0284**				
	(2.473)				
Depth	-0.0169***				
Buy-Side Submissions	(-3.779)	-0.0366***		-0.0318***	
Buy-side Submissions		(-5.605)		(-4.277)	
Buy-Side Executions		0.000979		0.0161*	
Day Side Executions		(0.125)		(1.649)	
Buy-Side Depth		-0.00973*		-0.0179***	
		(-1.889)		(-3.927)	
Sell-Side Submissions			-0.0418***		-0.0309***
			(-6.025)		(-4.542)
Sell-Side Executions			0.0127		0.0110
Call Cirla David			(1.621) -0.0105***		(1.171) -0.0104**
Sell-Side Depth			(-2.628)		(-2.309)
Observations	43,520	43,520	43,520	43,520	43,520
Number of Stocks	8	8	8	8	8
Stock FE	YES	YES	YES	YES	YES
Day FE	YES	YES	YES	YES	YES
Lagged Dep. Var	YES	YES	YES	YES	YES
Within R ²	0.352	0.283	0.283	0.264	0.264

- Market markers submit MPID orders when submissions and depth are low and when executions are high.
- ⇒ Supports Hypothesis H1.

Dep.Var	(1) MPID _t i	(2) MPID _t ; _{BUY}	(3) MPID _t ; _{BUY}	(4) MPID _t ;SELL	(5) MPID _t ;SELL
D	0.00405	0.000540	0.00000	0.00554*	0.00500*
Pricing Errors	-0.00105	0.000648	-0.00202	0.00651*	0.00690*
D: Cl	(-0.264)	(0.186)	(-0.602)	(1.720)	(1.806)
Price Changes	-0.0332***	-0.0340***	-0.0288***	-0.0431***	-0.0230**
wi:ΛPR<0	(-4.360)	(-4.305)	(-3.403)	(-5.367)	(-2.469)
$I_{t-1}^{i;\Delta PR < 0}$	0.00930	0.0117		-0.00827	
Price Changes $\times \mathbb{I}_{*-1}^{i;\Delta PR < 0}$	(0.705)	(0.902)		(-0.595)	
Price Changes ×III	0.0126	0.0156		0.0315**	
#i:ΛPR>0	(1.057)	(1.279)	0.0172	(2.341)	0.0007*
$I_{t-1}^{i;\Delta PR>0}$			-0.0173		0.0227*
D: CI #i:APR>0			(-1.378)		(1.690)
Price Changes $\times \mathbb{I}_{t-1}^{i;\Delta PR>0}$			0.00875		-0.0145
			(0.750)		(-1.146)
Observations	43,520	43,520	43,520	43,520	43,520
Number of Stocks	8	8	8	8	8
Stock FE	YES	YES	YES	YES	YES
Day FE	YES	YES	YES	YES	YES
Lagged Dep. Var	YES	YES	YES	YES	YES
Within R ²	0.352	0.283	0.283	0.264	0.264

- Market makers less likely to submit following price changes (and more likely to sell following negative price changes).
- ⇒ Less evidence to support Hypothesis H1a.

Market Reactions to MPID Order Submissions

- Endogeneity problem between the market maker's submission decision and ex-post market conditions.
- Solution: Heckman (1979) correction model.
- First Stage: Probit model to obtain Heckman selectivity correction term

$$\begin{aligned} y_t^i &= \alpha_0^i + \delta \mathbf{X}_{t-1}^i + \varepsilon_t^i, \\ \mathbb{I}_t^{i;MPID} &= \begin{cases} 1 & \text{if} \quad y_t^i > 0 \\ 0 & \text{otherwise}, \end{cases} \end{aligned}$$

- $\mathbb{I}_{t}^{i;MPID} = 1$ if $MPID_{t}^{i}$ is in the upper stock-day quartile.
- **X** $_{t-1}^{i}$: control variables.
- Heckman selectivity correction term uses predicted values:

$$\lambda_t^i = \frac{\phi(-\hat{y}_t^i)}{1 - \Phi(-\hat{y}_t^i)}.$$

 \bullet ϕ , Φ : standard normal density and cumulative distribution functions.

■ Second Stage:

$$q_t^i = \alpha_0^i + \alpha_1^i \lambda_{t-1}^i + \gamma \mathbb{I}_{t-1}^{i; MPID} + \beta' \mathbf{x}_{t-1}^i + \gamma' \mathbf{m}_{t-1}^i + \delta' \mathit{DAY}_t + \varepsilon_t^i$$

- λ_{t-1}^i : Heckman correction term.
- \mathbf{q}_t^i : ex-post market quality measure.
- Market Quality Measures:
 - Relative Bid-Ask Spreads, volatility, pricing errors, price changes.
 - (Buy- and sell-side) Submissions, executions, and depth.

Market Reactions to MPID Order Submissions

(1) RELSPR _t	$(2) VOL_t^i$	(3) PR.ERR ⁱ
0.0322*** (3.620)	0.00753 (0.717)	0.0191* (1.786)
0.398	0.329	0.147
43,520	43,520	43,520
8	8	8
YES	YES	YES
YES	YES	YES
YES	YES	YES
	0.0322*** (3.620) 0.398 43,520 8 YES YES	RELSPRi VOLi 0.0322*** 0.00753 (3.620) (0.717) 0.398 0.329 43,520 43,520 8 8 YES YES YES YES YES YES

- \blacksquare Spreads increase, no decrease in volatility. \Rightarrow No support for Hypothesis H2a.
- Increase in pricing errors. \Rightarrow No support for Hypothesis H2b.
- Market maker MPID orders do not stabilize markets.

Dep.Var	$SUB_t^{i;BUY}$	(6) $SUB_t^{i;BUY}$	(7) SUB _t ;SELL	(8) SUB _t ;SELL
$\mathbb{I}_{t-1}^{i; MPID; BUY}$	0.00887 (0.919)		-0.0407*** (-4.269)	
$\mathbb{I}_{t-1}^{i;MPID;SELL}$,	-0.0563*** (-6.308)	,	-0.0291*** (-2.986)
Within R ²	0.372	0.368	0.343	0.340
Observations	43,520	43,520	43,520	43,520
Number of Stocks	8	8	8	8
Stock FE	YES	YES	YES	YES
Day FE	YES	YES	YES	YES
Lagged Dep. Var	YES	YES	YES	YES

- Submission volumes decrease (particularly on the opposite side of the book) by about \$300.000-\$500.000.
- ⇒ No support for Hypothesis 2.

Dep.Var	$EXE_t^{i;BUY}$	$(10) \\ EXE_t^{i;BUY}$	$EXE_t^{i;SELL}$	(12) EXE ^{i;SELL}
$\mathbb{I}_{t-1}^{j;MPID;BUY}$	-0.0187* (-1.943)		-0.00820 (-0.900)	
$\mathbb{I}_{t-1}^{i;MPID;SELL}$,	-0.00973 (-1.029)	,	-0.0342*** (-3.463)
Within R ²	0.272	0.259	0.219	0.236
Observations	43,520	43,520	43,520	43,520
Number of Stocks	8	8	8	8
Stock FE	YES	YES	YES	YES
Day FE	YES	YES	YES	YES
Lagged Dep. Var	YES	YES	YES	YES

- Execution volumes decrease (particularly on the same side of the book) by about \$25,000-\$60,000.
- ⇒ No support for Hypothesis 2.

Market Reactions to MPID Order Submissions

	(13)	(14)	(15)	(16)
Dep.Var	$DEPTH_t^{i;BUY}$	DEPTH _t ;BUY	$DEPTH_t^{i;SELL}$	DEPTH _t ;SELL
$\mathbb{I}_{t-1}^{i;MPID;BUY}$	0.00176		-0.0385***	
1-1	(0.187)		(-4.534)	
$\mathbb{I}_{t-1}^{i;MPID;SELL}$,	-0.0353***	,	0.00957
1-1		(-4.203)		(0.917)
Within R ²	0.390	0.389	0.314	0.319
Observations	43,520	43,520	43,520	43,520
Number of Stocks	8	8	8	8
Stock FE	YES	YES	YES	YES
Day FE	YES	YES	YES	YES
Lagged Dep. Var	YES	YES	YES	YES

- Depth decreases (particularly on the opposite side of the book) by about \$6.000-\$9.000.
- ⇒ No support for Hypothesis 2.
- Market maker MPID orders do not encourage market participation.

Robustness Checks

- ✓ Shorter time intervals (30-seconds).
- ✓ Running regressions individually on a stock-by-stock basis.
- ✓ Only consider submissions from Timber Hill, LLC (market maker that accounts for 76% of total MPID submission volume in our sample).

Summary

- Examines whether MPID-attributed orders by participants in Nasdag's QMM program are aligned to the market maker role.
- Results show that market makers for the most part respond to market conditions in which liquidity provision is needed.
 - High spreads, high volatility, low submissions, low depth, high executions.
- However, less evidence that they submit contrarian orders.
- Therefore, little evidence that the non-anonymous orders submitted by QMMs help to improve market quality (and may even have negative effects).
- Exchanges may do well to extend liquidity incentive programs beyond the provision of limit orders.
 - E.g., NYSE Designated Market Maker program explicitly require participants to maintain price continuity and intervene when volatility is high.

References I

- Anand, A. and Venkataraman, K. (2016). Market conditions, fragility and the economics of market making. Journal of Financial Economics, 121(2):327-349.
- Benhami, K. (2006). Liquidity providers valuation of anonymity: The Nasdaq market makers evidence. Working Paper.
- Bessembinder, H., Hao, J., and Zheng, K. (2015). Market making contracts, firm value, and the ipo decision. The Journal of Finance, 70(5):1997-2028.
- Biais, B., Declerck, F., and Moinas, S. (2017). Who supplies liquidity, how and when? Working Paper.
- Chordia, T., Roll, R., and Subrahmanyam, A. (2008). Liquidity and market efficiency. Journal of Financial Economics, 87(2):249-268.
- Chung, K. H. and Chuwonganant, C. (2014). Uncertainty, market structure, and liquidity. Journal of Financial Economics, 113(3):476-499.
- Clapham, B., Gomber, P., Lausen, J., and Panz, S. (2017). Liquidity provider incentives in fragmented securities markets. Working Paper.
- Comerton-Forde, C., Putnins, T. J., and Tang, K. M. (2011). Why do traders choose to trade anonymously? Journal of Financial and Quantitative Analysis, 46(04):1025–1049.
- Demsetz, H. (1968). The cost of transacting. The Quarterly Journal of Economics, pages 33-53.
- Garbade, K. D. and Silber, W. L. (1979). Structural organization of secondary markets: Clearing frequency. dealer activity and liquidity risk. The Journal of Finance, 34(3):577-593.
- Grossman, S. J. and Miller, M. H. (1988). Liquidity and market structure. The Journal of Finance, 43(3):617-633.
- Harris, L. (1997). Order exposure and parasitic traders. Working Paper.

- Hasbrouck, J. (1993). Assessing the quality of a security market: A new approach to transaction-cost measurement. Review of Financial Studies, 6(1):191-212.
- Heckman, J. (1979). Sample specification bias as a selection error. Econometrica, 47(1):153-162.
- Hendershott, T. and Menkveld, A. J. (2014). Price pressures. Journal of Financial Economics, 114(3):405 423.
- Karam, A. (2018). Dealers' incentives to reveal their names. Working Paper.
- Malinova, K. and Park, A. (2017). Does high frequency trading add noise to prices? Working Paper.
- Panayides, M. A. (2007). Affirmative obligations and market making with inventory. Journal of Financial Economics, 86(2):513-542.
- Van Kervel, V. and Menkveld, A. J. (2019). High-frequency trading around large institutional orders. The Journal of Finance, 74(3):1091-1137.
- Vayanos, D. (2001). Strategic trading in a dynamic noisy market. The Journal of Finance, 56(1):131-171.
- Watanabe, M. (2017). Supply ambiguity and market fragility. Working Paper.