# Order placement behavior in a high frequency environment

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## The rise of algorithmic trading on Indian equity markets



Order placement behavior in a high frequency environment

#### What we know about AT

- Studies in the context of NSE equity markets:
  - Boehmer and Shankar (2013) find that increase in AT does not result in high levels of systemic risk.
  - 2 Aggarwal and Thomas (2014) find that
    - higher AT improves market liquidity, reduces volatility and increase price efficiency.

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  - Boehmer and Shankar (2013) find that increase in AT does not result in high levels of systemic risk.
  - Aggarwal and Thomas (2014) find that
    - higher AT improves market liquidity, reduces volatility and increase price efficiency.
    - **2** does not increase the incidence of extreme price movements.
    - (a) key: improvement in market quality of small market cap stocks.
  - Aggarwal et al (2017) examine the impact of a regulatory fee (to curb high orders to trades ratio) on market quality.
    - Find that when the fee was imposed by the exchange, it had its intended consequence of bringing down the OTR, and improve market quality.
    - Not so when the fee was imposed by the regulator.

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  - Q.5 If yes, do these cancellation occur from inside the touch?

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- NSE equity markets, an interesting case:
  - U.S: a fragmented market structure.
  - Data: long time series of data that can be used to distinguish a *low* AT and *high* AT period, & with every order and trade tagged as AT and non AT.

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Using this data, we recreate the full order book.

- Segment: NSE spot and stock futures.
- Period: Nov Dec 2009 (Prior co-location) and Nov Dec 2013 (post co-location).
- Sample: Top 200 firms by market cap in 2009 and 2013.
- Final sample: 147 stocks.

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# Q.1: Are there differences in the type of orders submitted by AT and non AT?

# Type of orders (spot)

MCap quartiles	Q1 (Highest)		Q4 (Lowest)		
	2009	2013	2009`	2013	
Algo orders (%)	18.61	78.95	5.93	41.47	
Fraction of algo or	ders (%)				
IOC	0.69	6.41	0.95	4.29	
SL	0.32	0.02	0.57	0.10	
MO	5.91	1.21	12.05	6.41	
Hidden	9.65	10.83	11.38	21.22	
Limit	83.43	81.53	75.05	67.98	
Fraction of non-algo orders (%)					
IOC	0.07	0.32	0.21	0.34	
SL	2.21	2.71	0.82	1.12	
MO	7.22	7.57	5.40	4.13	
Hidden	10.09	9.16	17.93	8.81	
Limit	80.41	80.24	75.64	85.60	
MCap (Rs. Mn)	843,191	1,030,189	51,885	62,322	
AT Intensity (%)	23.56	78.40	10.42	39.74	
# of orders	63,300	112,901	10,415	19,898	
# of stocks	37	37	37	37	

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# Type of orders (SSF)

MCap quartiles	Q1 (Highest)		Q4 (Lowest)			
	2009	2013	2009	2013		
Algo orders (%)	39.82	93.42	28.45	80.62		
Fraction of algo or	ders (%)					
IOC	7.16	9.40	4.38	11.10		
SL	0.00	0.02	0.01	0.09		
MO	0.48	0.15	0.14	0.15		
Spread	6.10	8.03	4.43	9.69		
Limit	86.26	82.41	91.05	78.96		
Fraction of <b>non-algo</b> orders (%)						
IOC	4.33	1.32	2.01	0.54		
SL	0.63	2.56	0.65	1.58		
MO	1.80	4.07	0.66	1.86		
Spread	4.85	3.26	2.80	1.71		
Limit	88.39	88.80	93.87	94.31		
AT Intensity (%)	22.42	73.46	11.77	47.18		
# of orders	64,743	226,195	8,744	36,163		
# of stocks	34	37	17	20		

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# Q.2 What is the pattern of activity on AT and non AT orders?

#### Order last activity by AT and non AT (Spot)

	As % of orders entered			
MCap quartiles	Q1 (Highest)		Q4 (Lowest)	
	2009	2013	2009	2013
Algo	17.96	78.41	3.81	38.66
Entered	0.15	0.03	0.17	0.10
Modified	0.02	0.02	0.01	0.04
Traded (A)	5.36	14.84	1.78	9.52
Cancelled (B)	12.43	63.52	1.84	29.00
NonAlgo	80.64	20.77	93.69	58.03
Entered	7.45	2.54	17.77	9.29
Modified	0.59	0.21	1.11	0.70
Traded (C)	50.11	14.31	49.44	35.84
Cancelled (D)	22.48	3.71	25.38	12.19
Cancelled (B+D)	34.92	67.23	27.23	41.19
Traded (A+C)	55.47	29.15	51.22	45.36
# of orders	59,043	102,768	9,026	16,848
# of stocks	37	37	37	37

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### Order last activity by AT and non AT (SSF)

	As % of orders entered			
MCap quartiles	Q1 (Highest)		Q4 (Lowest)	
	2009	2013	2009	2013
Algo	39.30	93.19	26.98	78.41
Entered	0.02	0.03	0.03	0.11
Modified	0.04	0.08	0.06	0.17
Traded (E)	1.48	2.17	0.90	2.50
Cancelled (F)	37.77	90.91	26.00	75.63
NonAlgo	59.90	6.40	71.29	20.86
Entered	1.30	0.46	2.66	1.63
Modified	0.33	0.10	0.74	0.32
Traded (G)	13.12	2.69	15.29	9.43
Cancelled (H)	45.16	3.15	52.60	9.49
Cancelled (F+H)	82.93	94.06	78.60	85.12
Traded (E+G)	14.59	4.85	16.19	11.93
# of orders	58,140	192,537	7,340	27,506
# of stocks	34	37	17	20

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# Q.3 Within the trades, do AT supply liquidity or demand it?

# Liquidity provisioning by AT and non AT (Spot)



# Liquidity provisioning by AT and non AT (SSF)



## Intraday liquidity demand and supply by AT (Q1)



## Intraday liquidity demand and supply by AT (Q4)



### Q4: What is the speed of order cancellations?

#### Cancelled algo orders: Spot



Q4 (Lowest market cap



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#### Cancelled algo orders: SSF







# Q.5: At what level of depth do cancellations occur?

#### Spot: Orders cancellations with duration less than a second

As % of orders cancelled in <1 second					
MCap quartiles	Q1 (Highest)		Q4 (Lowest)		
Order location at	2009	2013	2009	2013	
Entry, Exit					
As	% of orders	cancelled	in <1 s		
(< 1], (< 1]	13.40	5.31	51.97	39.01	
(1, 3], (< 1]	0.37	0.98	0.49	1.66	
(3, 5], (< 1]	0.02	0.39	0.00	0.12	
(> 5], (< 1]	0.01	0.03	0.02	0.01	
(< 1], (1, 3]	1.18	2.52	1.77	8.49	
(1, 3], (1, 3]	7.30	10.29	13.26	20.97	
(3, 5], (1, 3]	0.46	1.88	0.29	0.85	
(> 5], (1, 3]	0.05	0.14	0.05	0.03	
(< 1], (3, 5]	0.08	0.10	0.03	0.10	
(1, 3], (3, 5]	1.07	1.34	0.37	0.90	
(3, 5], (3, 5]	5.83	12.84	4.44	11.02	
(> 5], (3, 5]	0.53	0.88	0.15	0.37	
(< 1], (> 5]	0.02	0.03	0.00	0.03	
(1, 3], (> 5]	0.12	0.11	0.01	0.22	
(3, 5], (> 5]	0.94	1.79	0.18	0.85	
(> 5], (> 5]	68.64	61.35	26.95	15.39	
		As % of a	II cancelle	d orders	
'Fast'	5.74	54.80	1.19	30.60	
Algo 'fast'	3.63	54.49	0.43	29.66	

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#### SSF: Orders cancellations with duration less than a second

As % of orders cancelled in <1 second					
MCap quartiles	Q1 (Highest)		Q4 (Lowest)		
Order location at	2009	2013	2009	2013	
Entry, Exit					
	As % of orders cancelled in <1 s				
(< 1], (< 1]	10.63	4.66	32.43	12.06	
(1, 3], (< 1]	0.94	0.57	1.03	0.84	
(3, 5], (< 1]	0.11	0.10	0.09	0.12	
(> 5], (< 1]	0.15	0.02	0.06	0.02	
(< 1], (1, 3]	7.33	4.39	10.74	7.77	
(1, 3], (1, 3]	17.86	6.38	27.95	18.93	
(3, 5], (1, 3]	0.72	0.84	0.31	0.89	
(> 5], (1, 3]	0.30	0.06	0.07	0.05	
(< 1], (3, 5]	0.68	0.16	0.45	0.78	
(1, 3], (3, 5]	3.06	5.28	1.22	11.42	
(3, 5], (3, 5]	4.47	7.65	3.88	12.27	
(> 5], (3, 5]	0.89	0.66	0.33	0.67	
(< 1], (> 5]	7.20	5.97	6.38	6.24	
(1, 3], (> 5]	7.89	4.64	4.40	7.21	
(3, 5], (> 5]	4.14	4.61	1.62	5.82	
(> 5], (> 5]	33.64	54.02	9.05	14.92	
	As % of all cancelled orders				
'Fast'	9.02	74.48	3.17	53.14	
Algo 'fast'	4.90	73.21	1.19	52.47	

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#### In summary,

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- A:2 AT contribute significantly to order cancellations (55% on spot market & 90.23% on SSF),

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- **A.3** AT supply almost as much liquidity as the demand.
- **A.4** Significant percentage of order cancellations within one second (44% on spot and 63% on SSF).

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- **A.3** AT supply almost as much liquidity as the demand.
- **A.4** Significant percentage of order cancellations within one second (44% on spot and 63% on SSF).
- A.5 But most of these 'fast' cancellations away from the touch.

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### **Going forward**

- What explains high rate of order cancellations?
- Do we see higher cancellations during stress periods

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- What explains high rate of order cancellations?
- Do we see higher cancellations during stress periods
- How could these characteristics be explained in terms of the underlying variables of market liquidity?

Thank you

#### Comments / Questions?

http://www.ifrogs.org/

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