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The Anatomy of Trading Algorithms

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How do people trade?

- Retail
 - Internalization and price improvement
- HFT
 - High frequency market making
 - "Arbitrage"
- Institutions (the "buy-side")

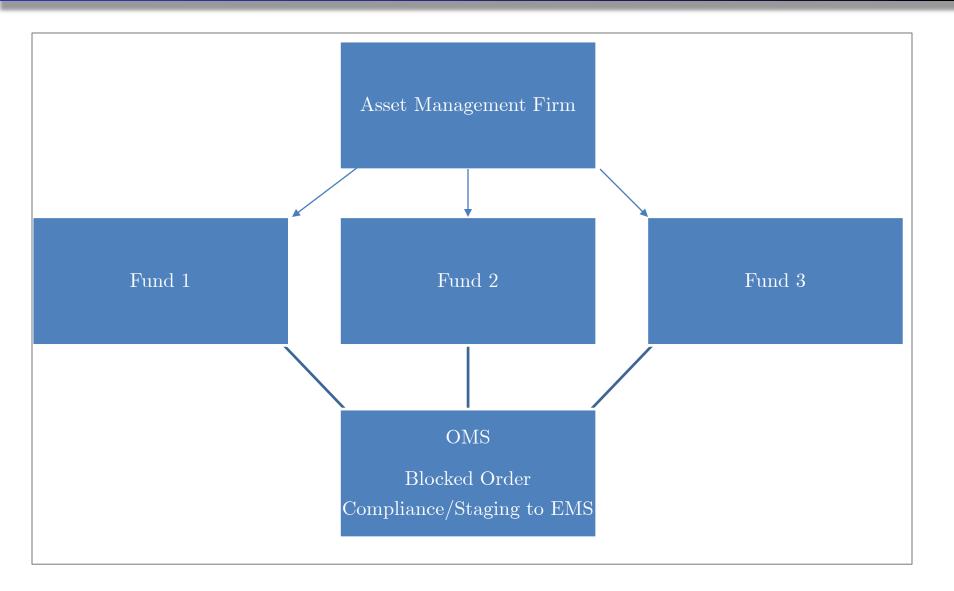
Background

- Classical models are outdated
- New models focus on HFT, ignoring buy-side
- Why?
 - No straightforward intuition (exception, LWY)
 - No data
 - Institutions don't want to release the data
 - Brokerage firms don't want to release the data
 - Ancerno-type data are essentially useless
 - Single-manager datasets: embedded endogenous choices, bespoke trading, generalizability issues

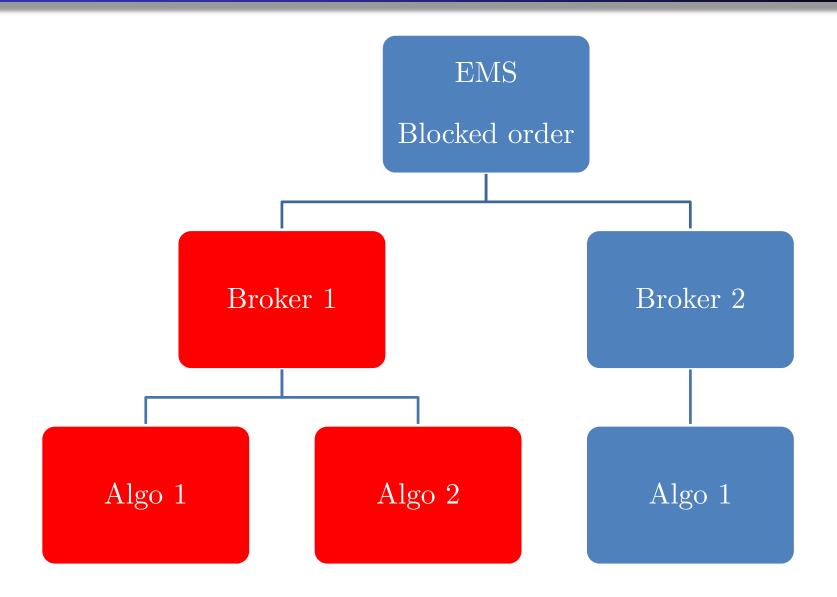
Objectives

- 1. Anatomy
 - To inform price formation
 - Anatomy informs diagnosis, and if necessary, cure
 - Gross anatomy (Parent)
 - Cytology (Child)
 - Histology (Runs)
- 2. To study the tradeoff between likelihood of execution and transaction costs
 - Fundamental issue
 - Building block of order choice models and price formation

Trading Process: Level 1 (OMS)



Trading Process: Level 2 (EMS)



Data overview

- Large provider of algorithmic trading services
 - Diverse client range, from buy-side long-only managers to multi-asset hedge funds. 961 clients, 2012-2016
- Four standardized, non-bespoke single-stock algorithms.
 - Use direct exchange feeds, often white-labelled
 - All time stamps in milliseconds with FIX tags
- \$675 billion in demand, trading over 5,000 securities
- Parent -> child ->fill
- 2.3 million parent orders generate
 - 300 million child submissions
 - \$2.1 trillion in notional volume
 - \$388 billion in traded volume

Data elements

Parent

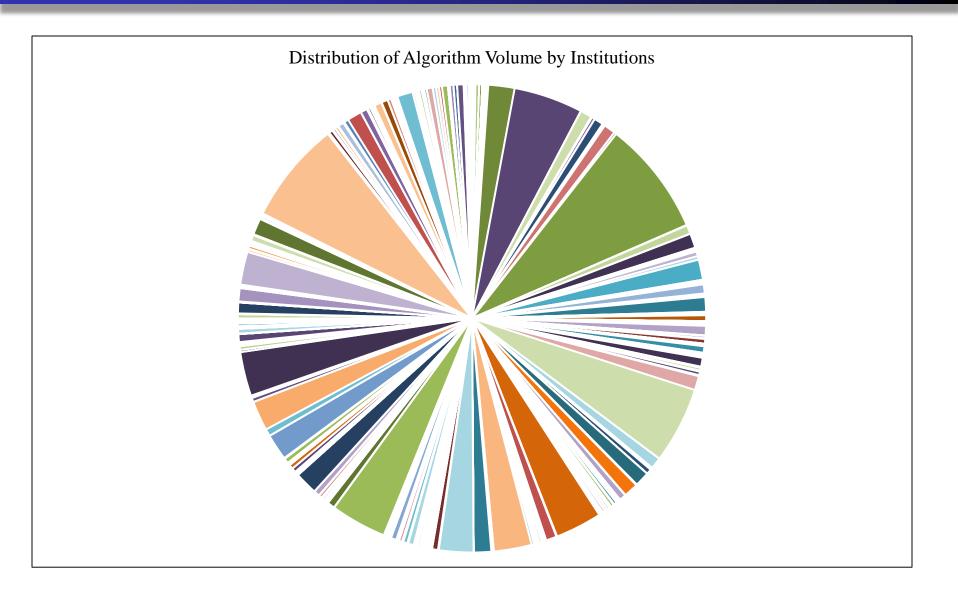
- Client ID
- Algo type
- ID, Symbol, side (B, S, SS)
- Start/End time
- Order Quantity
- Price and/or volume constraints

Child (all, not just filled)

- ID, link to parent
- Time (send, receive)
- Order type (M, L, P)
 - Subtype (limit prices)
- Display instruction
- Exec Instr. (FIX 18)
- TIF (FIX 59)
- Venue
- Fill (Price, Quantity)
- Add/Take/Route (FIX 851)
- Fee/Rebate

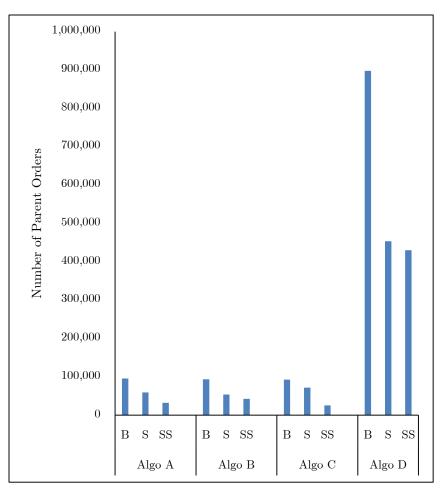
Gross Anatomy: Parent Orders

Dollar distribution of algo usage

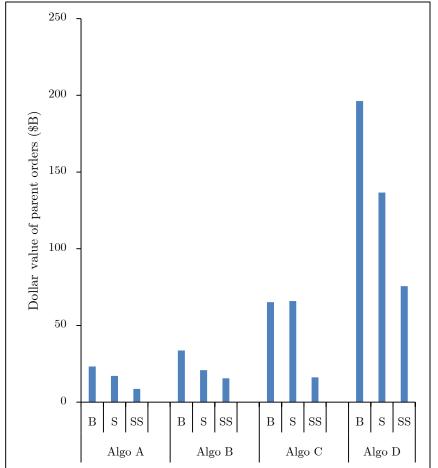


Distribution of parent orders

By Number

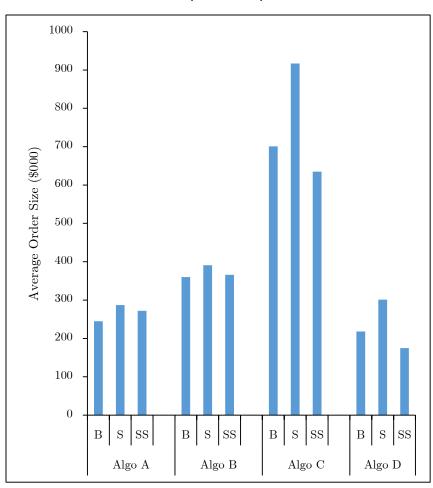


By Dollar Volume (\$B)

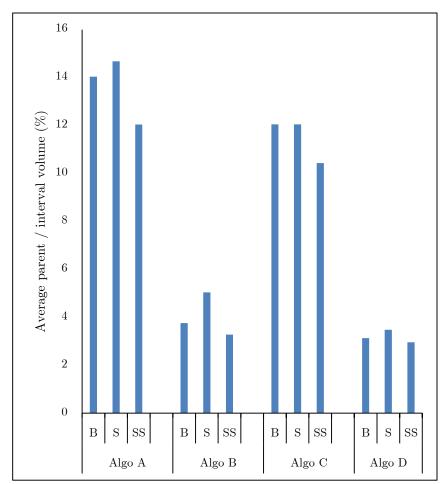


Average parent size

Parent size (\$000)

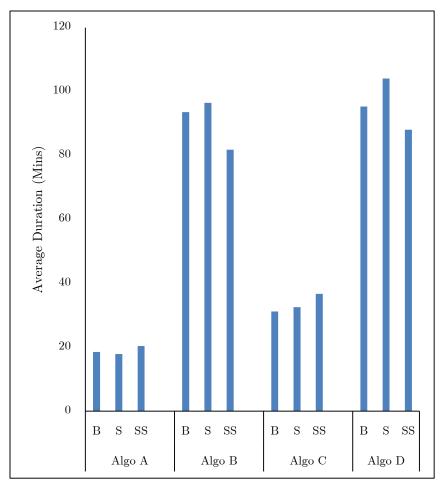


Parent / interval volume

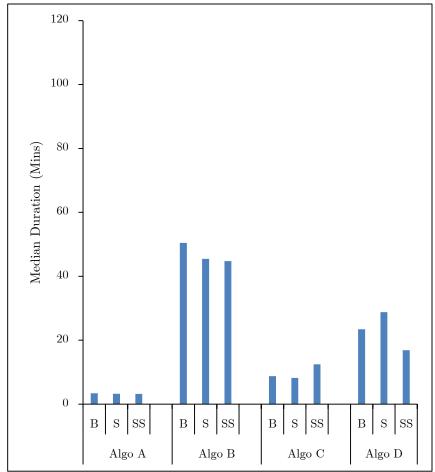


Parent duration

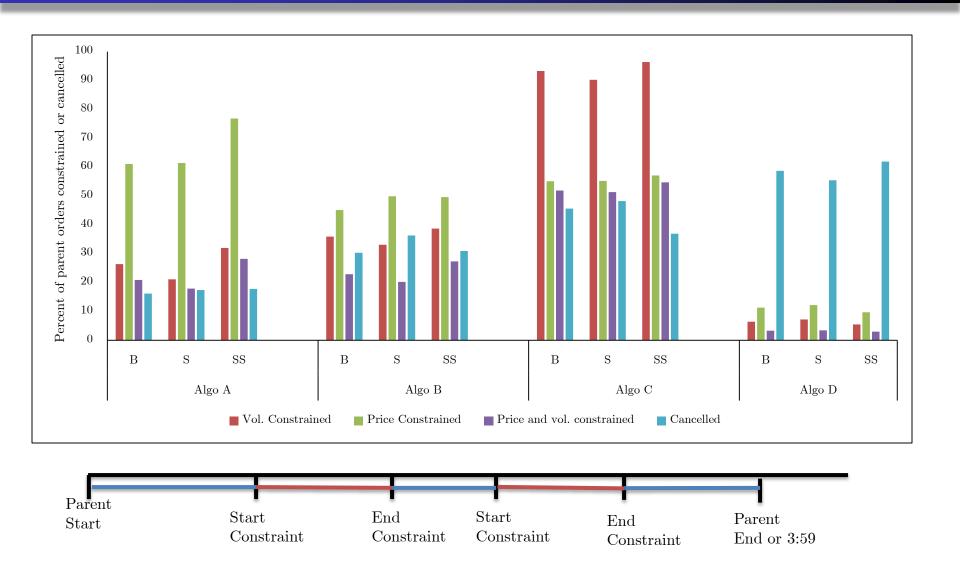
Mean



Median

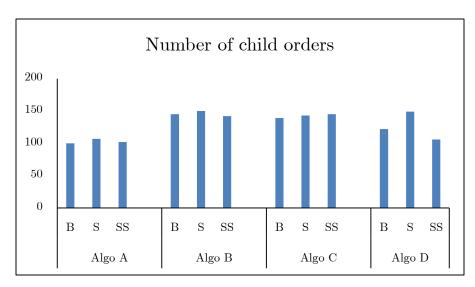


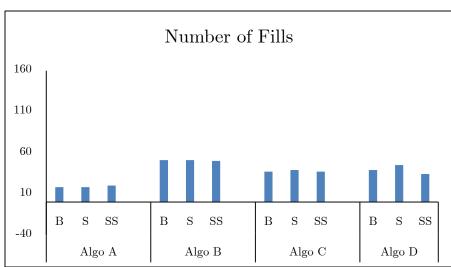
Parent constraints & cancellations

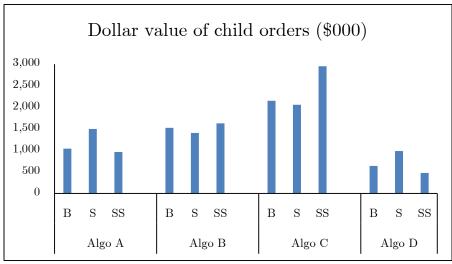


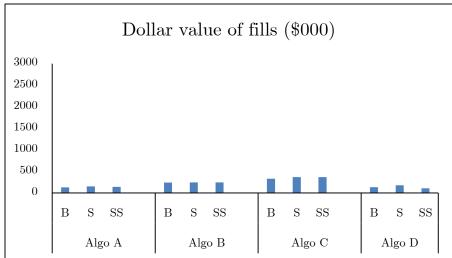
Cytology: Child Orders

Child orders and fills



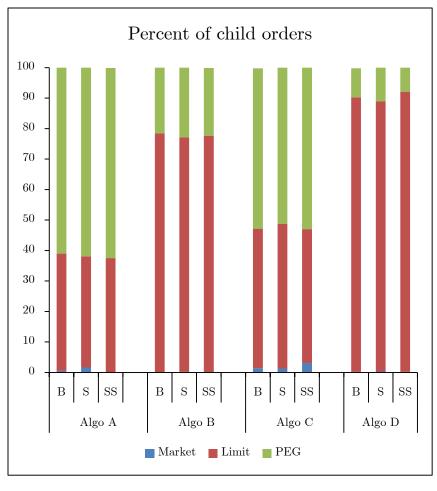




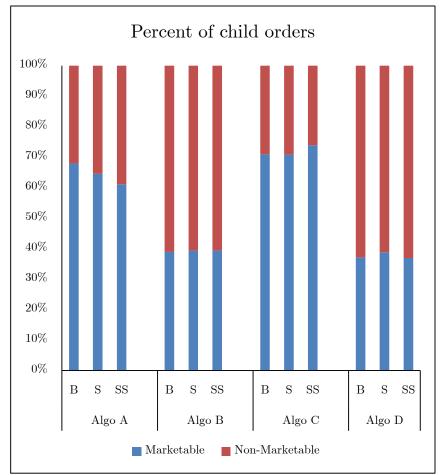


Child order characteristics

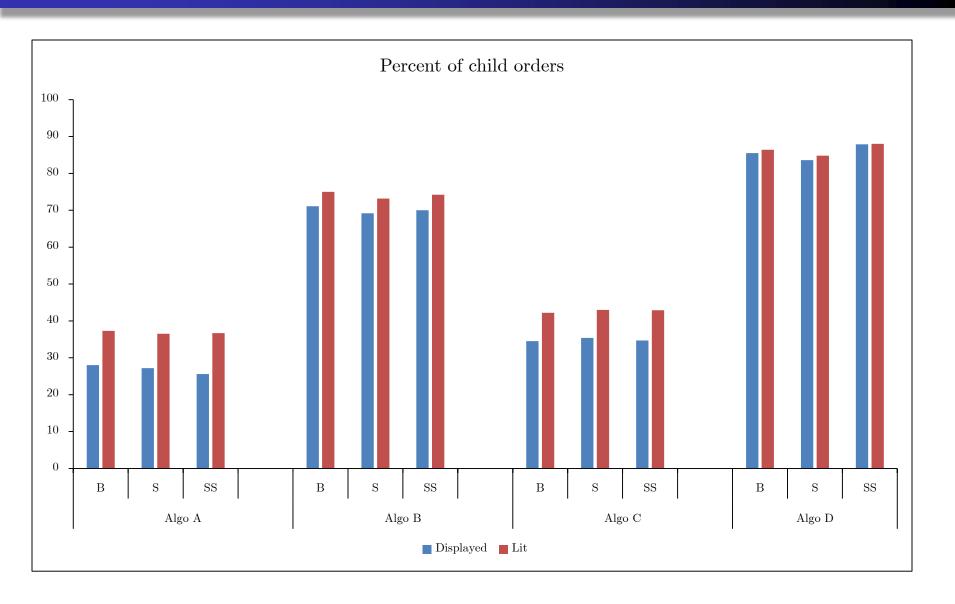
Order type



Marketability



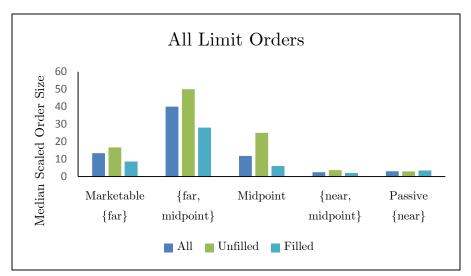
Display and Venue

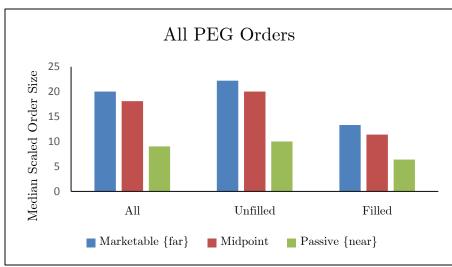


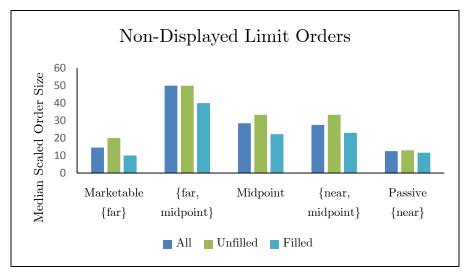
Choices and Outcomes

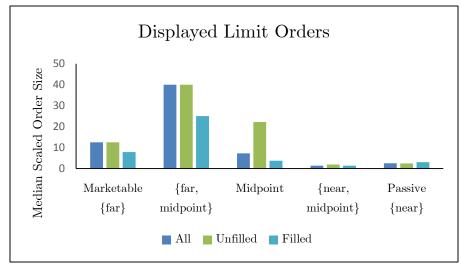
- 1. Choices
 - a) Price
 - b) Size
 - c) Display
- 2. Outcomes: Execution Risk vs Transaction Costs
 - a) Fill Rates
 - b) Time to fill
 - c) Cost

Choices

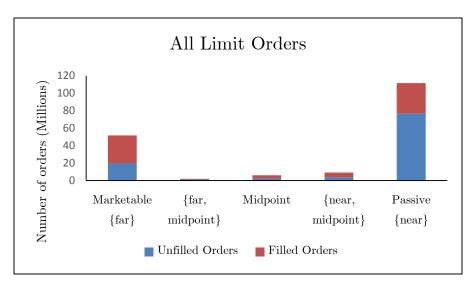


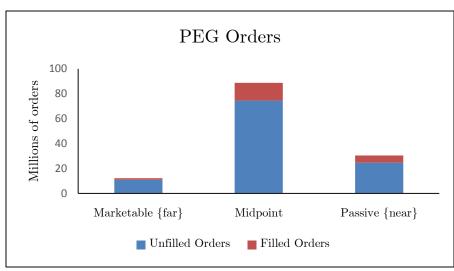


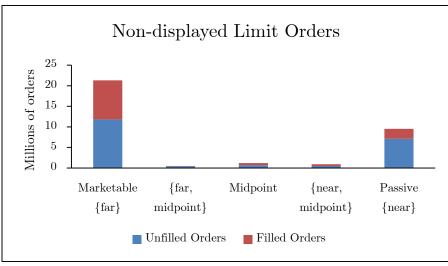


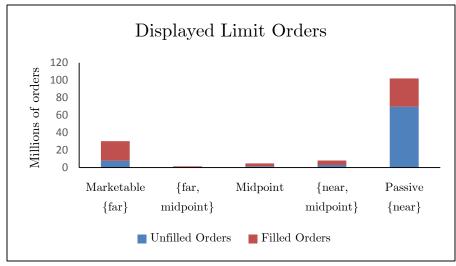


Outcome: Fill Rates





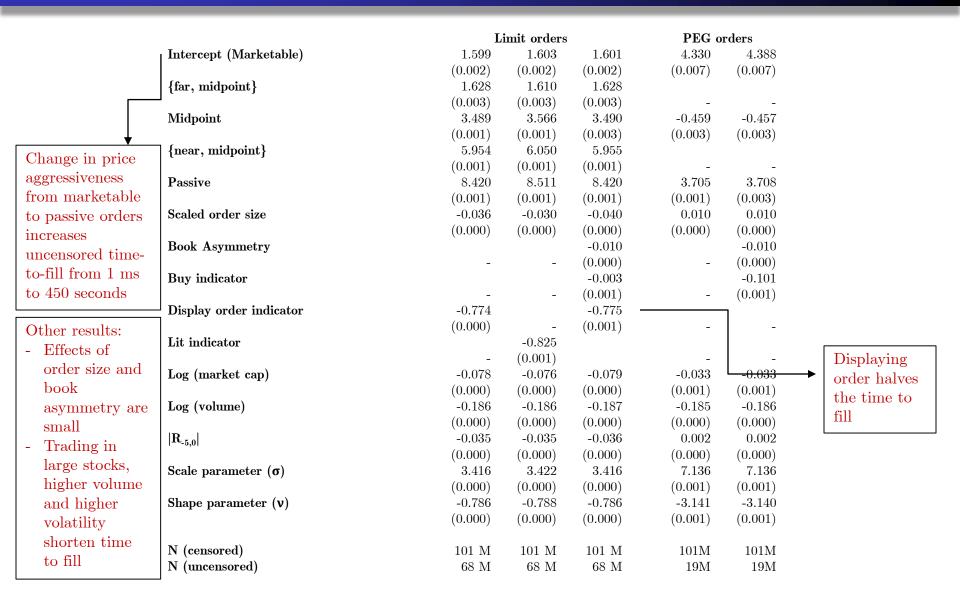




Outcome: time to fill

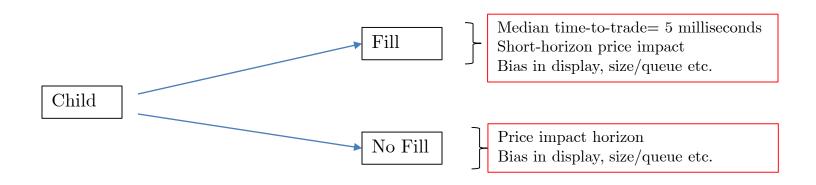
- For marketable and ex post filled orders:
 - Time-to-fill(Passive)>>Time-to-fill
- For marketable but ex post unfilled orders:
 - Time-to-fill ~ Time-to-cancel
- For passive orders, time to fill is about ½ of time to cancel
- This is useful, but
 - Cancellations affect the conditional distribution of time-to-fill, i.e. censoring.
 - And, execution risk is a function of endogenous choice variables and random market conditions
- Accelerated Failure Time (AFT) models (Lo, MacKinlay, Zhang (2002)). Model T (the life of the order).

Accelerated Failure Time models



Price Movements

- Canonical theory: **trades** incorporate private information and move prices
- Does expression of **trading interest** move prices?
- Passive quotes **supposedly** liquidity providing and do not contain private information
 - Brogaard, Hendershott, Riordan (2019)



Measurement

• Child price impact

$$cpi_{jt\tau} = q_{jt}(m_{j,t+\tau} - m_{jt}) / m_{jt}$$

- For child submissions, t is submission time.
- For fills, t is trade time.
- Horizon (τ) issues as in Conrad and Wahal (2020).
 - $\tau = 100 \text{ ms}, 500 \text{ ms}, 1 \text{ second}, 5 \text{ second}, 10 \text{ seconds}$ post submission/fill

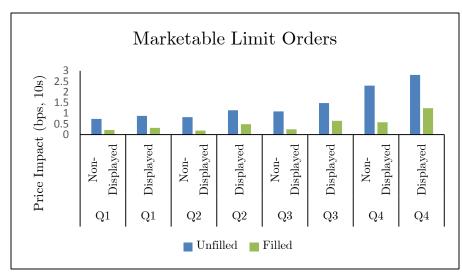
Price Impact

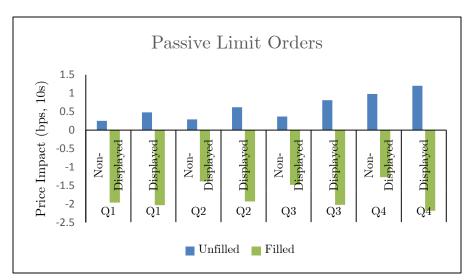
	Price Impact in bps					Price Impact in bps					
Submission Price	$100 \mathrm{ms}$	$500 \mathrm{ms}$	1 s	5 s	10 s	$100 \mathrm{ms}$	$500\mathrm{ms}$	1 s	5 s	10 s	
Aggressive (Far)	1.19	1.25	1.28	1.66	2.03	0.41	0.41	0.43	0.64	0.80	
$\{Far, Midpoint\}$	0.57	0.67	0.71	1.25	1.72	-0.12	-0.15	-0.16	0.09	0.26	
Midpoint	0.30	0.33	0.36	0.73	1.06	-0.43	-0.49	-0.55	-0.75	-0.83	
$\{Near, Midpoint\}$	0.67	0.77	0.85	1.43	1.97	-0.91	-1.08	-1.27	-1.75	-2.02	
Primary (Near)	0.08	0.14	0.18	0.51	0.84	-1.04	-1.23	-1.44	-1.82	-2.01	
Unfilled Child PEG Orders Price Impact in bps											
	U				s		Filled Ch Price 1	ild PEG Impact i			
Submission Price	100ms				s 10 s	$100 \mathrm{ms}$				10 s	
Submission Price Aggressive (Far)		Price I	mpact in	n bps			Price 1	Impact i	n bps	10 s 0.14	
	$100 \mathrm{ms}$	Price I 500ms	impact in	n bps 5 s	10 s	$100 \mathrm{ms}$	Price 1 500ms	Impact i	n bps 5 s		
	$100 \mathrm{ms}$	Price I 500ms	impact in	n bps 5 s	10 s	$100 \mathrm{ms}$	Price 1 500ms	Impact i	n bps 5 s		
Aggressive (Far)	100ms 0.25	Price I 500ms 0.51	mpact in 1 s 0.54	5 s 0.82	10 s 1.06	100ms -0.06	Price 3 500ms -0.04	Impact is 1 s -0.05	5 s 0.05	0.14	

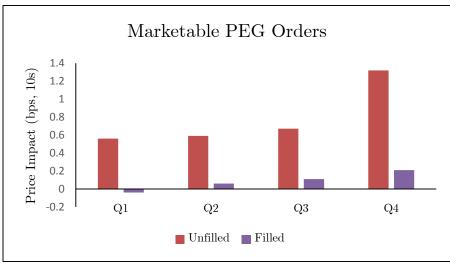
Unfilled Child Limit Orders

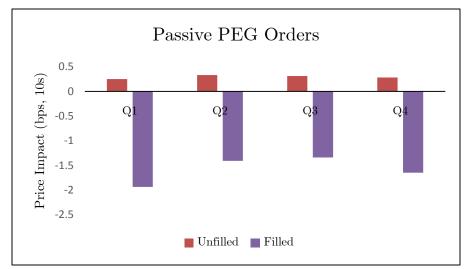
Filled Child Limit Orders

Price Impact: Triple Sort









Regressions: Limit Orders

	Unfille	ed	Filled			
Intercept	1.215	1.360	0.388	0.490		
${f (Marketable)}$	(0.017)	(0.028)	(0.008)	(0.009)		
$\{{ m far,\ midpoint}\}$	-0.634	-0.555	-0.646	-0.573		
	0.028)	(0.031)	(0.028)	(0.031)		
Midpoint	-0.894	-0.912	-1.404	-1.404		
	(0.019)	(0.020)	(0.050)	(0.032)		
$\{{ m near,midpoint}\}$	-0.355	-0.403	-2.372	-2.407		
	(0.025)	(0.025)	(0.025)	(0.026)		
Passive $\{near\}$	-1.110	-1.220	-2.458	-2.502		
	(0.016)	(0.020)	(0.008)	(0.026)		
Display	0.170	0.224	0.217	0.226		
	(0.010)	(0.018)	(0.008)	(0.007)		
Scaled size	0.099	· -	0.151	_		
	(0.004)		(0.006)			
Book Asymmetry*Buy	·	0.002	·	0.001		
		(0.000)		(0.000)		
Display * Scaled size	0.005	-	-0.037	-		
	(0.001)		(0.007)			
$ \mathrm{Ret}_{-5,0} $	0.104	0.117	-0.024	-0.017		
	(0.001)	(0.014)	(0.001)	(0.001)		
Buy	$0.02\acute{6}$	0.004	0.006	$\stackrel{\cdot}{0.001}$		
·	(0.012)	(0.014)	0.007)	(0.008)		
	()	, ,	,	,		
Average N	63,815	63,815	$45{,}125$	$45,\!125$		
Average adj- \mathbb{R}^2	0.103	0.076	0.110	0.104		

Regressions: PEG Orders

	Unfille	ed	Filled			
Intercept	0.581	0.688	0.093	0.110		
(Marketable)	(0.022)	0.026)	(0.015)	(0.016)		
{far, midpoint}	·	-	·	-		
Midpoint	-0.160	-0.199	-0.557	-0.549		
	(0.025)	(0.026)	(0.015)	(0.016)		
$\{ { m near, midpoint} \}$	-	- -	-	-		
Passive {near}	-0.508	-0.585	-1.357	-1.660		
	(0.022)	(0.025)	(0.019)	(0.019)		
Scaled size	0.060	-	0.038	_		
	(0.003)		(0.005)			
Book Asymmetry*Buy	_	0.003	·	0.001		
		(0.000)		(0.000)		
$ \mathrm{Ret}_{-5.0} $	0.058	0.063	-0.042	-0.041		
. 3,5,	(0.001)	(0.001)	(0.002)	(0.002)		
Buy	0.011	-0.002	0.006	0.015		
	(0.009)	(0.010)	(0.008)	(0.010)		
Average N	66,478	66,478	12,228	12,228		
${\bf Average \ adj\text{-}R^2}$	0.038	0.029	0.037	0.027		

Histology: Strategic Runs

Strategic Runs

- The sequence of child orders is not random (à la Hasbrouck and Saar (2013)
 - Common goal
 - Shared codebase
- Run: a sequence of child orders from the same parent in the same price aggressiveness category
- Collapse categories into:
 - Marketable
 - Inside
 - Passive
- Parents that seek at least 1 basis point of ADV and with at least 50 child orders

Description & Transition Matrices

	All Algos	$\mathbf{Algo} \mathbf{A}$	$\mathbf{Algo}\;\mathbf{B}$	$\mathbf{Algo} \mathbf{C}$	${\bf Algo~D}$
# parents	812,132	$50,\!582$	82,903	87,751	$590,\!896$
Runs/parent	63.09	53.59	51.65	41.23	68.76
Child per run	8.84	11.92	7.45	8.16	8.87
Run duration	566.99	102.04	452.81	158.58	683.46
Run volume	1515.26	5503.08	3024.40	3063.60	732.23
Percent runs					
Passive	45.58	40.58	42.78	42.00	46.94
Inside spread	16.04	16.34	19.98	10.94	16.21
Marketable	38.37	43.06	37.23	47.04	36.84

	All Algos		Algo A		Algo B		Algo C		Algo D)					
	$P_{\rm t}$	${ m I_t}$	$\mathrm{M_{t}}$	$ P_{ m t} $	${ m I_t}$	$\mathrm{M_{t}}$	$ P_{ m t} $	${ m I_t}$	$\mathrm{M_{t}}$	P_{t}	${f I}_{ m t}$	$M_{\rm t}$	$ P_{ m t} $	${ m I_t}$	M_{t}	
$\mathbf{P}_{ ext{t-1}}$	-	13.9	31.2	-	10.3	33.0	-	17.0	27.2	-	3.0	41.1	-	14.6	30.7	
$\mathbf{I}_{ ext{t-1}}$	12.9	-	5.2	9.2	-	7.1	13.7	-	6.6	2.8	-	6.9	13.9	-	4.8	
$\mathbf{M}_{ ext{t-1}}$	29.3	7.4	-	31.3	9.1	-	25.7	9.8	-	38.4	7.8	-	28.9	7.0	-	

Logistic Regressions

- Does the price aggressiveness of the run depend on
 - Whether the prior run received a fill $(Fill_{t-1})$
 - "Cost" of prior run: cumulative signed midpoint return (SRet_{t-1})
- Dependent variables: Prob(Marketable), Prob(Passive)

	Algo A		\mathbf{Alg}	o B	\mathbf{Alg}	o C	Algo D	
	$\mathrm{Fill}_{\mathrm{t-1}}$	$\mathrm{SRet}_{\mathrm{t-1}}$	$\mathrm{Fill}_{\mathrm{t-1}}$	$\mathrm{SRet}_{\mathrm{t-1}}$	$\mathrm{Fill}_{\mathrm{t-1}}$	$\mathrm{SRet}_{\mathrm{t-1}}$	$\mathrm{Fill}_{\mathrm{t-1}}$	$\mathrm{SRet}_{\mathrm{t-1}}$
Aggressiveness of prior run								
Passive	0.524	0.005	0.022	0.001	0.205	0.006	-0.041	0.001
	(0.005)	(0.000)	(0.003)	(0.000)	(0.006)	(0.000)	(0.001)	(0.000)
	[7.59]	[0.09]	[0.50]	[0.03]	[1.07]	[0.04]	[0.83]	[0.01]
Inside	-0.067	-0.017	-0.374	-0.006	-0.017	-0.018	-1.359	-0.010
	(0.006)	(0.000)	(0.004)	(0.000)	(0.007)	(0.000)	(0.001)	(0.000)
	[1.65]	[0.44]	[-8.32]	[0.14]	[-0.35]	[0.01]	[26.95]	[0.19]
Marketable	0.063	-0.007	0.098	-0.001	0.366	-0.014	0.659	-0.000
	(0.05)	(0.004)	(0.004)	(0.000)	(0.004)	(0.000)	(0.001)	(0.001)
	[0.92]	[-0.11]	[1.71]	[-0.02]	[5.06]	[-0.18]	[11.21]	[0.00]

Logistic Regressions

- Dependent variables: Prob(Marketable), Prob(Passive)
- Independent variables:
 - Whether the prior run received a fill
 - The "cost" of the prior run: signed midpoint return

	Algo A					Algo B Algo C				Algo D			
	$\Pr(M_t)$	$\mathrm{Pr}(\mathrm{M_t})$	$\Pr(P_t)$	$\Pr(M_t)$	$\Pr(M_t)$	$\Pr(P_t)$	$\Pr(\mathrm{M_t})$	$\Pr(M_t)$	$\Pr(P_t)$	$\Pr(\mathrm{M_t})$	$\mathrm{Pr}(\mathrm{M_t})$	$\Pr(P_{t})$	
$\mathbf{Run}_{\mathrm{t-1}}$	P	I	M	P	I	${ m M}$	P	I	${ m M}$	P	I	M	
$\mathbf{Fill}_{\mathbf{t-1}}$	0.52	-0.07	0.06	0.02	-0.37	0.09	0.20	-0.02	0.36	-0.04	-1.35	0.65	
	(0.00)	(0.01)	(0.05)	(0.00)	(0.00)	(0.00)	(0.01)	(0.01)	(0.00)	(0.00)	(0.00)	(0.00)	
	[7.6]	[1.7]	[0.9]	[0.5]	[-8.3]	[1.7]	[1.1]	[-0.3]	[5.1]	[0.8]	[26.9]	[11.2]	
$\mathbf{SR}_{ ext{t-1}}$	0.01	-0.02	-0.01	0.01	-0.01	-0.00	0.01	-0.01	-0.01	0.00	-0.01	-0.00	
	(0.00)	(0.00)	(0.00)	(0.00)	(0.00)	(0.00)	(0.00)	(0.00)	(0.00)	(0.00)	(0.00)	(0.00)	
	[0.1]	[0.4]	[-0.1]	[0.0]	[0.1]	[0.0]	[.04]	[0.0]	[-0.1]	[0.0]	[0.2]	[0.0]	

Conclusions

- Central tradeoff: trading versus incurring trading costs. Thus the use of algorithms.
- To do so, they generate hundreds of child orders
- Child orders employ price, time, display priority rules to navigate the tradeoff
- Passive orders have much larger execution risk but still incur substantial price impact
- Marketable orders do not guarantee execution and generate even larger price impact