

# **The Rise of Machines and Liquidity Patterns in Markets**

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# Key Points

- Patterns emerge before reasons for them become apparent, especially in complex systems such as financial markets
- Big Data has assumed a big role in emergence of machine intelligence: computers are not only running markets but playing a major role in data driven hypothesis **generation**
- What does liquidity “feel like” with machines running markets and why?
- What do regulators need in terms of technology capability to keep up with the complexity of financial markets?

# Big Data and Search Algorithms: The Grist for Machine Intelligence

Data Science is the study of the generalizable extraction of knowledge from data\*

The fire hose of data – quotes, cancellations, trades, etc is massive and getting larger – complexity has been on the rise: the analytically astute have an advantage

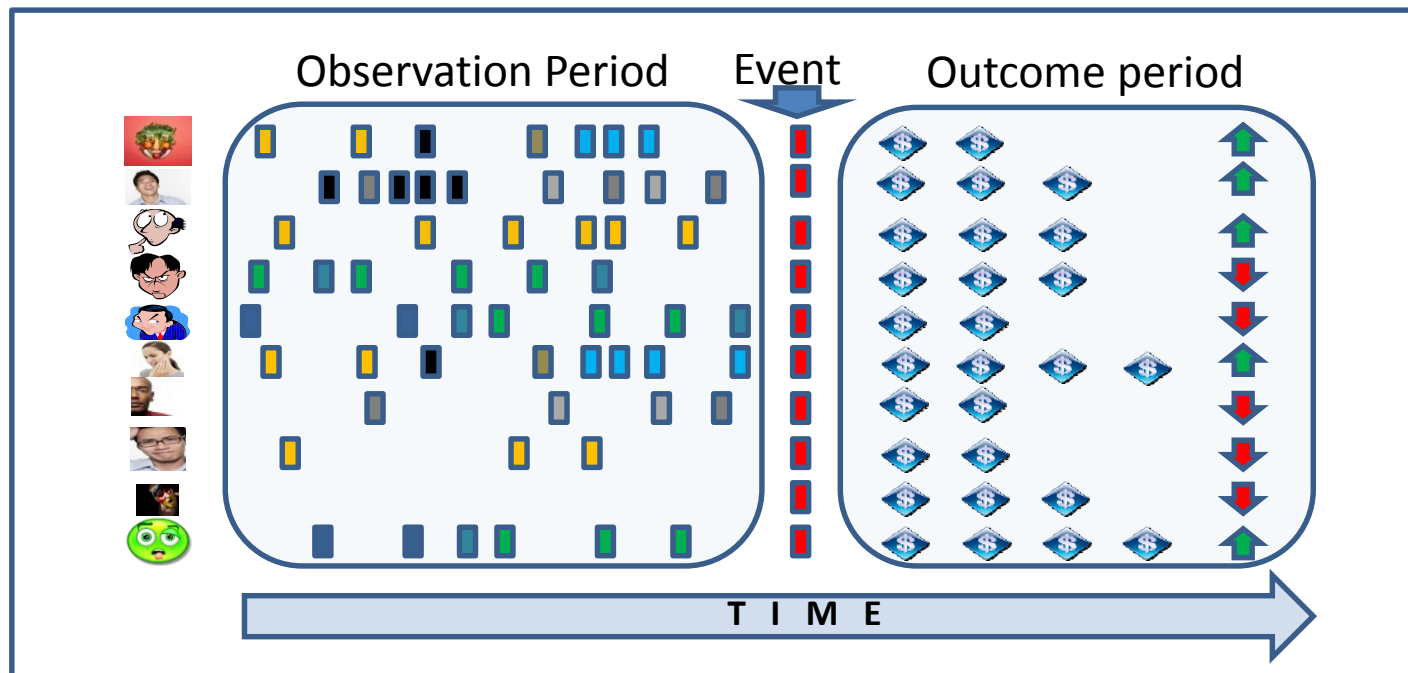
\*Dhar, V., Data Science and Prediction, Communications of the ACM, Vol. 56 No. 12, December 2013  
<http://cacm.acm.org/magazines/2013/12/169933-data-science-and-prediction/fulltext>

# Asking the Right Question

**“Patterns Emerge Before Reasons for Them Become Apparent”**

Asking the right question is therefore critical: “If only you knew what question to ask me, I’d give you very interesting answers from the data.”

# What is the Right Question Here?



- Are bad outcomes associated with the blues?
- Or with the grays?
- Or the yellows in the absence of the blues?
- Or is it more than three yellows or three blues?
- Or is it the greens in “quick succession?”
- Or ... do we need to run additional experiments and gather more data?

# Observational Data and Acquiring New Data

- Observational data may answer questions without explicitly asking anyone anything!
- It may also require an understanding of how the data are being generated
  - Are there “natural experiments” that are reflected in the data or are the data somehow biased through self selection?
  - Is it possible to **run** experiments to **get** additional data to amend the observed data?

# Pattern/Anomaly Detection Mindset...

...assume that most data are “noise”

...challenge is to extract the “signal” that occurs very infrequently

...design algorithms accordingly



# Internalization Patterns

Dark Pool / Internalization Rates by Volume and Stock Price – September 2012

		Consolidated Average Daily Volume							
		Under 100K CADV	100K-250K	250K-500K	500K-1MN	1MN-2MN	2MN-4MN	4MN-8MN	Above 8MN
Stock Price	Under \$1	36.6%	36.5%	40.0%	40.5%	38.1%	45.4%	42.7%	41.9%
	\$1-\$5	40.9%	44.2%	43.8%	43.4%	43.4%	40.7%	40.6%	46.9%
	\$5-\$10	39.7%	41.5%	39.6%	37.5%	38.6%	35.9%	31.3%	37.4%
	\$10-\$15	41.2%	40.3%	38.0%	36.8%	35.6%	31.0%	31.5%	35.6%
	\$15-\$20	43.4%	42.7%	38.8%	33.1%	35.7%	30.7%	29.7%	28.5%
	\$20-\$30	44.2%	42.8%	38.1%	34.5%	39.8%	32.3%	30.0%	34.7%
	\$30-\$40	40.9%	37.3%	37.3%	31.6%	30.9%	29.9%	32.0%	31.0%
	\$40-\$50	44.7%	34.2%	32.7%	33.5%	31.6%	30.8%	31.5%	27.8%
	\$50-\$75	46.0%	37.7%	33.2%	31.9%	31.0%	29.2%	29.3%	25.9%
	\$75-\$100	44.1%	38.5%	35.9%	31.5%	31.6%	30.5%	33.8%	28.4%
	\$100-\$150	41.3%	34.7%	34.2%	35.0%	33.9%	26.4%	26.7%	27.8%
	\$150-\$250	32.5%	29.2%	32.8%	28.9%	31.5%	27.8%	24.8%	33.8%
	\$250-\$400	24.3%	NA	NA	32.4%	NA	34.7%	NA	NA
	Above \$400	31.4%	NA	31.5%	32.2%	NA	35.4%	NA	37.0%

1. Is the penny spread artificially wide in these liquid stocks?

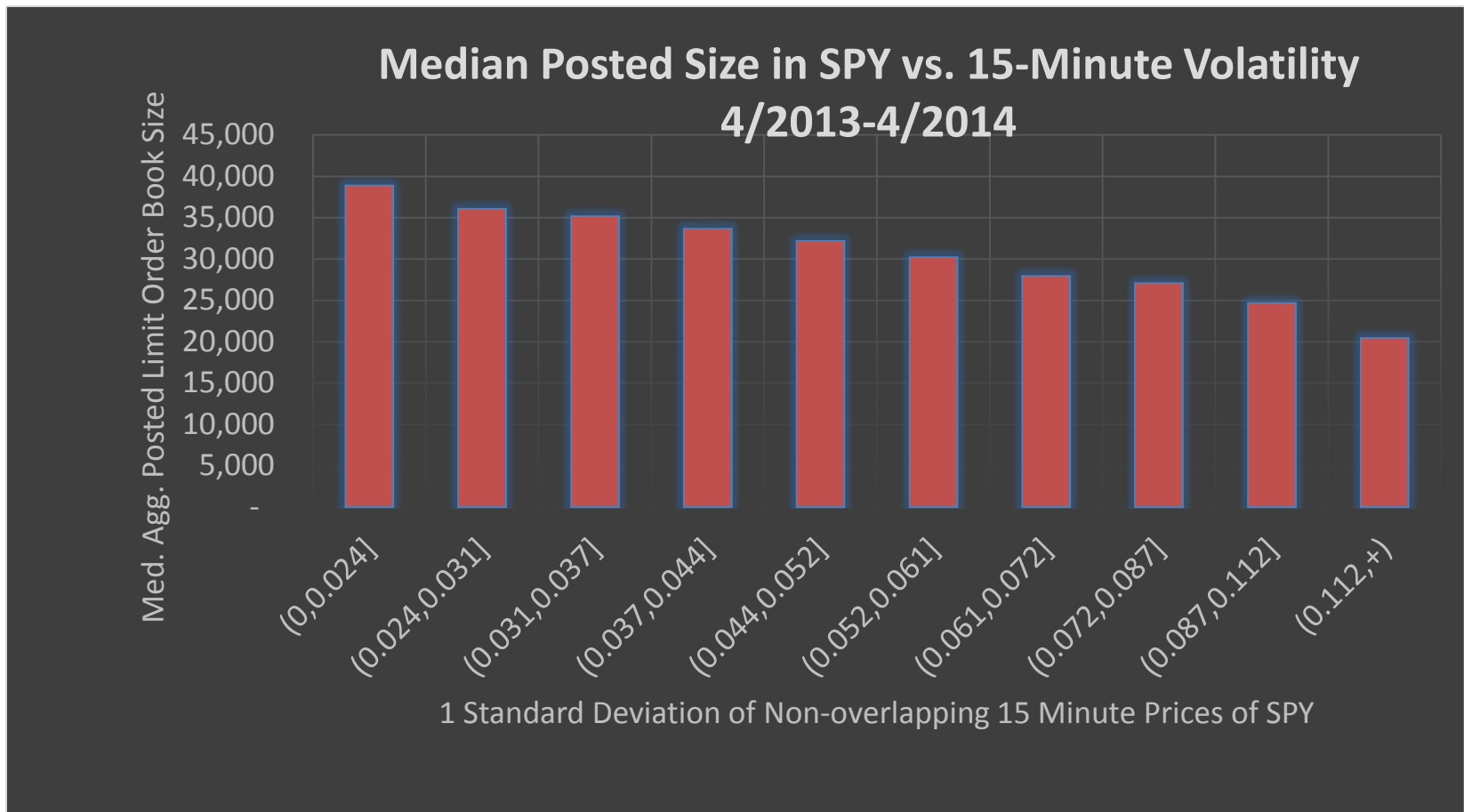
2. Do high-priced stocks require too much capital?

3. Is there insufficient liquidity to risk posting an order?

Source: TABB Group



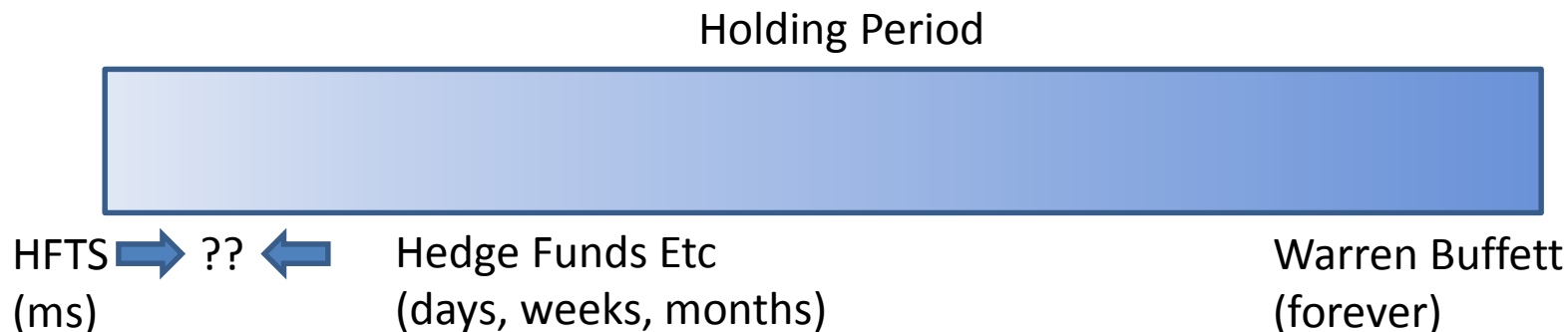
# Volatility Vs. Liquidity



# Is the “Displayed Liquidity” Illusory?

- It can be when you most need it, but the data suggest otherwise
  - But its always been that way to some extent even before machines
  - Perhaps the “cost of liquidation” is a better measure of liquidity: takes specific risk into account
- Making markets with penny spreads (as opposed to 6.25 cents in the old days) requires sophisticated real-time risk management
- *We wanted* a fast market and *encouraged fragmentation*, so it isn’t surprising that machines are doing more of the work to minimize risk and use every bit of information available to do so rapidly

# What if Speed Isn't an Advantage?



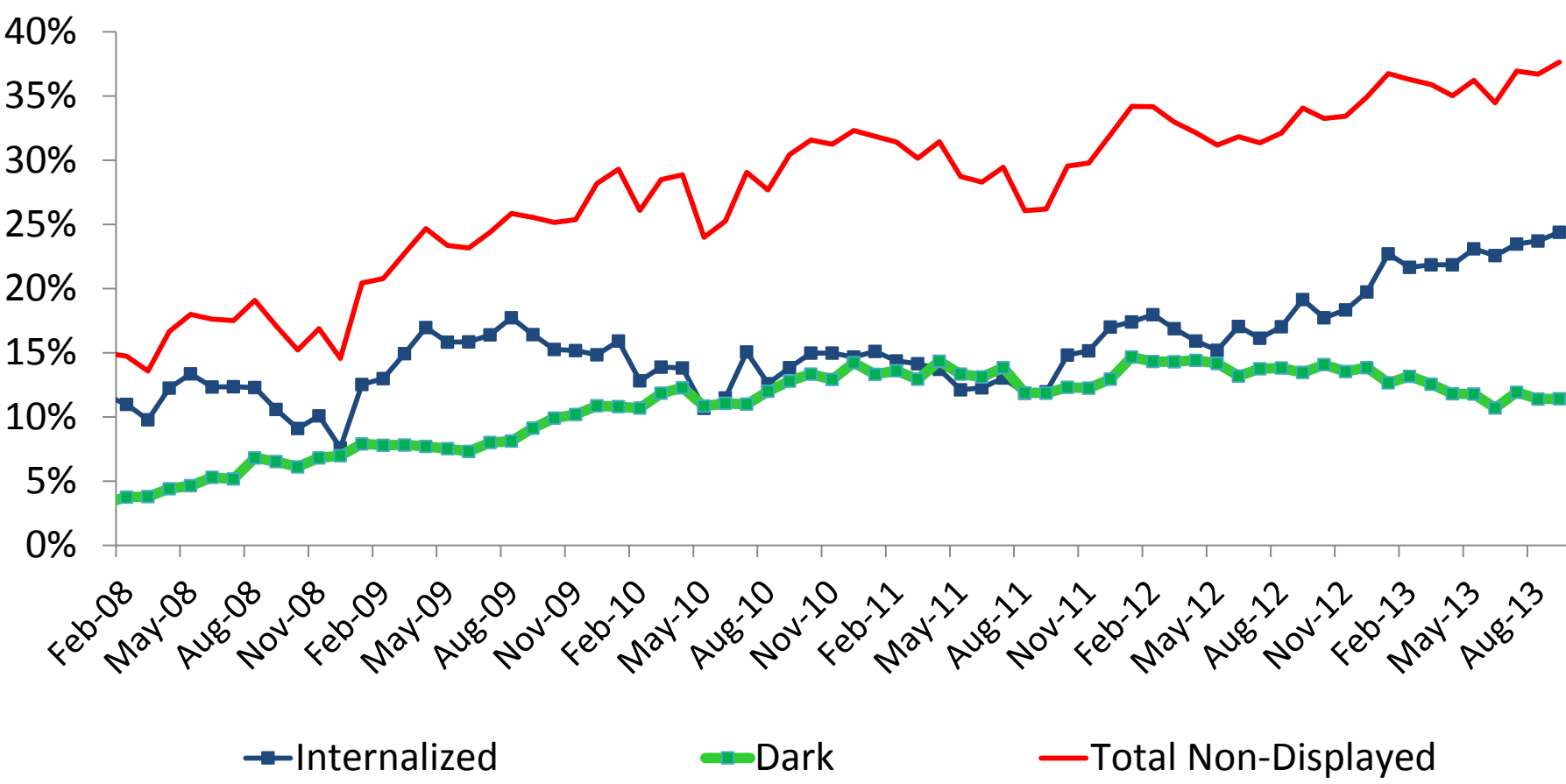
Is this one reason why HFT profitability is declining?

# Summary

- People “closer” to the market tend to have an advantage
- People who understand the implications of new regulation ask the right questions for discovering patterns in “big data” through better predictive analytics
- HFTs can not only be faster but smarter because they need to control risk tightly in current-day markets where spreads are tight
- If markets are “rigged” or “unfair” this should be evident in the data, as long as regulators have the analytical sophistication to detect patterns that suggest as such irregularities
- NMS had unintended consequences, but you can’t punish people for being smarter!



# Recent Data: Off-Exchange Activity Has Increased



Source: TABB Group