#### Overview of Electronic Market<sup>1</sup>

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<sup>&</sup>lt;sup>1</sup>Reference: Avellaneda (2011), Maglaras (2015)

#### Increasing Percentage of Algo Trading

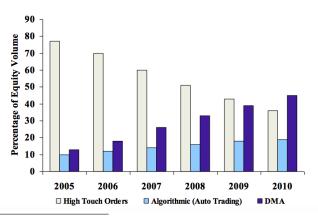


- A third of all European Union and United States stock trades in 2006 were driven by automatic programs, or algorithms.
- As of 2009, studies suggested HFT firms accounted for 60-73% of all US equity trading volume, with that number falling to  $\sim 50\%$  in 2012.
- ullet FX markets also have active algo trading ( $\sim 25\%$  in 2006).
- $\bullet$  Futures markets are considered fairly easy to integrate into algo trading, with  $\sim 20\%$  of options volume by 2010.

### Algo Trading in US Equity Market

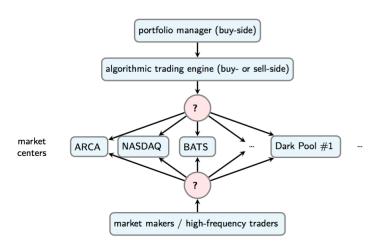
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### US Equities markets: percentage of orders generated by algorithms



<sup>&</sup>lt;sup>2</sup>Avellaneda (2011)

#### Simplified View of Trading



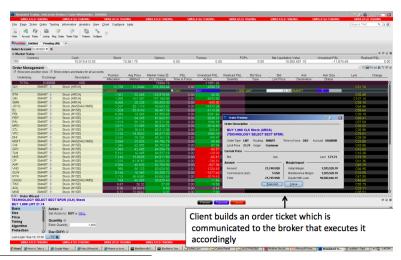
# A few trading venues for US equity markets<sup>3</sup>

- ARCA-NYSE: electronic platform of NYSE (ex- Archipelago)
- BATS: (Kansas)
- BEX: Boston Equity Exchange
- CBSX: CBOE Stock Exchange
- CSXZ: Chicago Stock Exchange
- DRCTEDGE: Direct Edge (Jersey City, NJ)
- ISE: International Securities Exchange
- ISLAND: Acquired by Nasdaq in 2003
- LAVA: belongs to Citigroup
- NSX: National Stock Exchange (Chicago)
- NYSE: New York Stock Exchange
- TRACKECN: Track ECN

<sup>&</sup>lt;sup>3</sup>Avellaneda (2011)

# Electronic order-management and execution system (client-broker)

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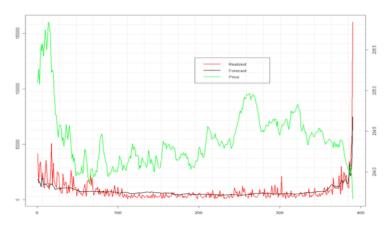
#### Modern US Equity Markets

- Electronic order-management and execution
- Decentralized/Fragmented
  - NYSE, NASDAQ, ARCA, BATS, Direct Edge, ...
- Exchanges ( $\sim$ 70%)
  - electronic limit order books (LOBs)
- Alternative venues (~30%)
  - ECNs, dark pools, internalization, OTC market makers, etc.
- Participants increasingly automated
  - institutional investors: algorithmic trading
  - market makers: high-frequency trading ( ${\sim}60\%$  average daily volume(ADV))
  - opportunistic/active (price sensitive) investors: aggressive/electronic
  - retail: manual (~5% ADV; small order sizes)

#### An Example

- How should you buy 250,000 shares of IBM stock between 12:30pm and 4:00pm?
  - Is this order "large"?
  - How fast should you trade? When to post orders?
  - How much will it cost you?
  - Who are you trading against?
- How is it done in practice?

#### Example Cont'd



- Forecasted Volume 12:30-4pm = 1,525,000 shares
  - Average spread = \$.04 (1.95bps)
  - $\bullet$  Expected Market Impact (12:30-4pm)  $\approx$  20bps  $\approx$  40 pennies/share
  - $\bullet$  Expected Market Impact (12:30-1:30pm)  $\approx$  28bps  $\approx$  56 pennies/share

#### Institutional Traders (Informed Traders)

- Institutional traders are usually informed traders.
- Investment decisions and trade execution are often separate processes.
- Institutional order flow typically has "mandate" to execute.
- Traders select brokers, algorithms, block venue, ...
  - (algorithm  $\approx$  optimization under trading constraints)
- Main considerations:
  - Best execution
  - Access to liquidity (larger orders)
  - Short-term alpha (discretionary investors)
  - Information leakage (large orders the spread over hours, days, weeks)
  - Commissions (soft dollar agreements)
  - Incentives (portfolio manager & trading desk; buy side & sell side)

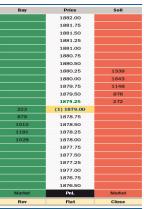
### Institutional Traders (Cont'd)

- Execution costs feedback affects:
  - Portfolio selection decisions
  - Hedge fund performance
- S&P 500:
  - ADV  $\approx < 1\%$
  - Market capital  $\approx .1\% \sim 2\%$
  - **Depth** (displayed, top of book)  $\approx$  .1% ADV
  - **Depth** (displayed, top of book)  $\approx 10^{-6} \sim 10^{-5}$  of MktCap
- Orders need to be spread out over time.

#### Market Depth

#### CLIMBING THE MARKET

A price ladder or DOM displays market depth data.



#### **CHARTING DEPTH**

Market depth as an overlay on a price chart. The green bars represent interested buyers; the red bars show interested sellers.



Source: www.baranalyzer.com

### Market Makers & HFT Participants

- Supply short-term liquidity and capture bid-ask spread
  - Mostly intraday flow
  - Limited overnight exposure
  - Small order sizes & depth
  - Short trade horizons/ holding periods
- ullet Profit pprox (Captured spread) (Adverse selection) (Trading cost)
- It is critical to model adverse selection.
  - Definition: Short-term price change conditional on a trade.
  - Essentially "uninformedness" price from information asymmetry.

## Market Makers & HFT Participants (Cont'd)

- It is important to model short-term future prices ("alpha")
  - Microstructure signals (limit order book & instant price impact)
  - Time series modeling of prices (momentum Versus reversion)
  - Cross-asset signals (statistical arbitrage, ETF against underlying, ...)
  - News (NLP)
  - Detailed microstructure of market mechanism (human psycho reaction)
  - ...
- Position risks:
  - Adverse price movements
  - Flow toxicity
  - Accumulation of inventory & aggregate market exposure

#### Toxic Flows

- Types of Toxic Flows <sup>5</sup>
  - Latency arbitrage or "picking-off" the feed
    - Trading on pricing engines of MM's that are slow in updating prices.
    - Slow-price reaction from inefficient technology/ unsophisticated model
  - Market impact of multiple orders
  - News Trading
- VPIN<sup>6</sup>: Volume-Synchronized Probability of Informed Trading, a measure of order toxicity.
  - Higher VPIN indicates that it is more likely that short-term momentum is due to informed trading.

<sup>&</sup>lt;sup>5</sup>Aratovskaya (2016)

<sup>&</sup>lt;sup>6</sup>Easley et al. (2012)

### Algo Trading Systems: Typically Decomposed into 3 Steps

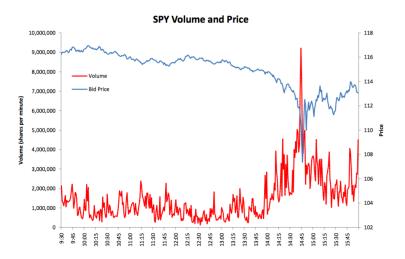
- Trade scheduling (macro-trader): splits parent order into  $\sim$  5 min slices (Lecture 2)
  - Relevant time-scale: minutes-hours
  - Schedule follows user selected strategy (VWAP, POV, IS, ... )
  - Reflects urgency, alpha, risk/return tradeoff
  - Schedule updated during execution to reflect price/liquidity/...
- Optimal execution of a slice (micro-trader): further divides slice into child orders (Lecture 3)
  - Relevant time-scale: secondsminutes
  - Strategy optimizes pricing and placing of orders in the LOB
  - Execution adjusts to speed of LOB dynamics, price momentum, ...
- Order routing: decides where to send each child order (Lecture 4)
  - Relevant time-scale:  $\sim 1-50$  ms
  - Optimizes fee/rebate tradeoff, liquidity/price, latency, etc

#### Algorithmic Trading Systems: basic building blocks

- forecasts for intra-day trading patterns
  - volume
  - volatility
  - bid-ask spread
  - **–** ...
- real-time market data analytics
- market impact model
- risk model
  - of the shelf risk models calibrated using EOD closing price data do not incorporate intra-day correlation structure
  - intra-day data? (tractable for liquid securities, e.g., S&P500 universe)
  - cross-asset liquidity model & market impact model

## The May 6, 2010 Flash Crash

7



<sup>&</sup>lt;sup>7</sup>Source: Joint CTFC SEC Report, 9/30/2010

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